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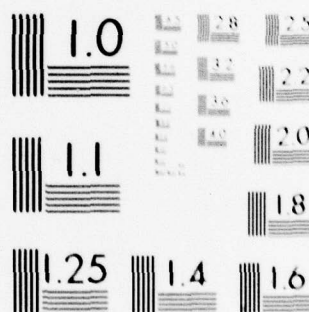
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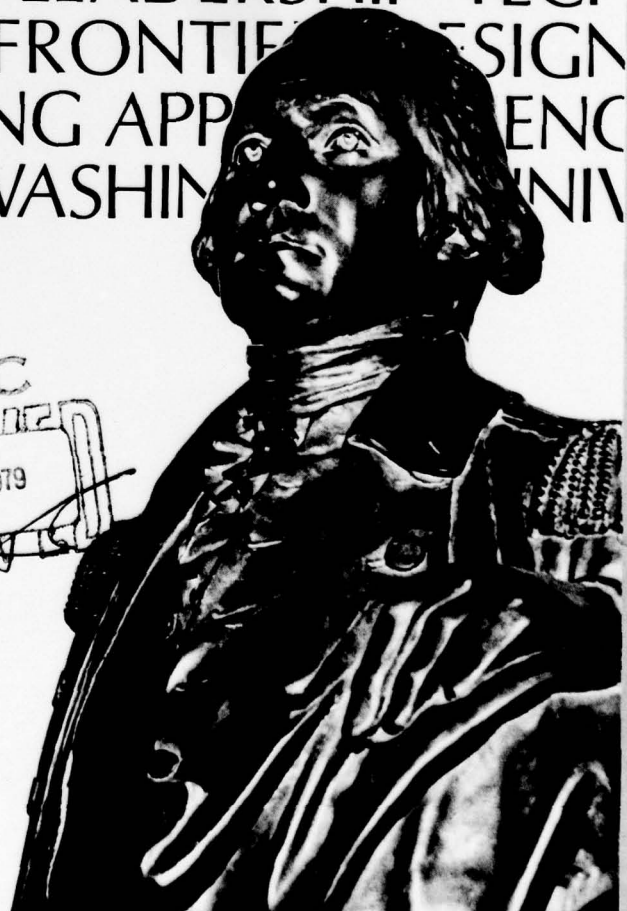
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# LEVEL II

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SYMBOLIC FACTORABLE SUMT: WHAT IS IT.  
HOW IS IT USED,

by

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Abolfazl Ghaemi  
Garth P. McCormick

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Model [1] is elaborated. The procedure for putting problems into symbolic factorable language that is digestible by the model, and the related model outputs, are fully explored. For illustrative purposes three increasingly difficult sample problems are coded, solved, and discussed in sufficient detail. For completeness, the theory of symbolic factorable sensitivity analysis that was developed in [4] is included as an appendix.

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## TABLE OF CONTENTS

LIST OF TABLES . . . . .	iii
LIST OF ILLUSTRATIONS . . . . .	iv
ABSTRACT . . . . .	v
1. INTRODUCTION . . . . .	1
2. FACTORABLE FUNCTIONS . . . . .	3
3. INPUT SPECIFICATIONS . . . . .	10
4. CODED EXAMPLES AND OUTPUT ILLUSTRATIONS . . . . .	21
4.1 Sample Problem 1 (A Test Problem) . . . . .	23
4.1.1 Problem definition . . . . .	23
4.1.2 Symbolic factorable form . . . . .	23
4.1.3 The keypunch sheet code . . . . .	24
4.1.4 Output annotations . . . . .	24
4.2 Sample Problem 2 (Hopkins Problem) . . . . .	32
4.2.1 Problem definition . . . . .	32
4.2.2 Computer listing of the symbolic code . . . . .	35
4.2.3 Selected pages from computer output . . . . .	35
4.3 Sample Problem 3 (Corrugated Transverse Bulkhead Design) . . . . .	35
4.3.1 Problem definition . . . . .	38
4.3.2 Computer listing of the symbolic code . . . . .	39
4.3.3 Selected pages from computer output . . . . .	39
5. GENERAL DESCRIPTION OF THE NEW SUBROUTINES . . . . .	43
REFERENCES . . . . .	47
APPENDIX 1 Annotated Computer Output for Solution of Sample Problem 1 (pages 50-68) and Selected Pages of Computer Solution Output for Sample Problems 2 (pages 69-77) and 3 (pages 78-84) . . . . .	49
APPENDIX 2 "Sensitivity Analysis in Nonlinear Programming Using Factorable Symbolic Input" . . . . .	85

## LIST OF TABLES

1. Example of Factorable Function (Example i) . . . . .	8
2. The SUMT Parameter Card . . . . .	11
3. The First Option Card . . . . .	12
4. The SYMPUT Parameter Card . . . . .	15
5. The Variable Definition Cards . . . . .	16
6. The Problem Parameter Cards . . . . .	17
7. The Separable and Quadratic Cards . . . . .	19
8. The Bound Cards . . . . .	20
9. The Tolerance Card . . . . .	20
10. The Second Option Card . . . . .	21
11. The Correspondence Between Functions of Sample Problem 1 in Symbolic Factorable Form and its Symbolic Factorable Code . . .	28
12. List of Single Variable Functions Used by SYMBOLIC FACTORABLE SUMT . . . . .	33
13. SUMT-Version 4 versus SYMBOLIC FACTORABLE SUMT on Solution of Optimal Corrugated Transverse Bulkhead Design Problem . . . . .	35
14. Problem Design Parameters (Sample Problem 3) . . . . .	40

## LIST OF ILLUSTRATIONS

1. Data Deck Structure . . . . .	22
2. Data Cards for Sample Problem 1 . . . . .	25
3. Computer Listing of Sample Problem 1 . . . . .	27
4. Computer Listing of Sample Problem 2 . . . . .	36
5. Computer Listing of Sample Problem 3 . . . . .	41



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This report is intended to serve as a guide for coding, solving, and conducting sensitivity and elasticity analysis on a solution vector of general parametric nonlinear programming problems using the latest version of the computer programming model of SYMBOLIC FACTORABLE SUMT. The motivation for development of this model, which uses symbolic factorable input, and the process of its evolution is briefly reviewed. The relative merits and advantages of this model over its earlier versions, as well as over the SUMT-Version 4 model [1] is elaborated. The procedure for putting problems into symbolic factorable language that is digestible by the model, and the related model outputs, are fully explored. For illustrative purposes three increasingly difficult sample problems are coded, solved, and discussed in sufficient detail. For completeness, the theory of symbolic factorable sensitivity analysis that was developed in [4] is included as an appendix.

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1. Introduction

This report is intended to guide individuals working in the area of optimization in the use of SYMBOLIC FACTORABLE SUMT, a computer programming model, to code, solve, and conduct sensitivity analysis studies on parametric mathematical optimization problems of the following general structure:

$$\begin{aligned} &\text{minimize } F(x,y) \\ &\quad x \in E^n \\ &\text{subject to } G(x,y) \geq 0 \\ &\quad H(x,y) = 0, \end{aligned}$$

where

$$\begin{aligned} G: E^n \times E^K &\rightarrow E^m \text{ and } C^2 \\ H: E^n \times E^K &\rightarrow E^p \text{ and } C^2 \end{aligned}$$

and  $y \in E^K$  corresponds to the problem parameters.

Since its development, SUMT-Version 4 [1] has undergone a series of modifications and refinements; SYMBOLIC FACTORABLE SUMT is its latest

version. The primary incentive for such modifications has been to relax the coding of the problem functions and their first and second derivatives, required by SUMT-Version 4 as users' subroutines, which in many cases have proved to be cumbersome, time consuming, costly, and prone to various derivational and coding errors. Table 13, provided in Section 4 for sample example 3, supports this claim.

The computer program reported here is an outgrowth of the program developed by McCormick [2]. The rationale for such an attempt was the fact that

"Almost every nonlinear function can systematically be decomposed to (and hence synthesized from) simple functions of one variable."

Using this fact, which in essence constitutes the backbone structure of factorable programming--the topic of the next section--McCormick developed a series of subroutines which internally calculated the values and the first and second derivatives of single variable functions, many of which are listed in Table 12. Each of these functions was associated with a numeric code, called a transformation code, which aided their identification in the routines. Another subroutine was provided to synthesize any desired combination of these single variable functions to reconstruct the intended problem functions. Interface of these routines with those of SUMT-Version 4 relaxed the provision of the user's subroutines to the model. In fact, the user's task boiled down to decomposition of his problem functions to functions of single variables and earmarking them with appropriate transformation codes. This was a major breakthrough which to a great extent alleviated the user's task in coding and solving optimization problems, and brought about significant savings of the time and cost involved in these steps.

Yet another attempt to facilitate the coding of problems was by deSilva and McCormick [3], through which evolved the original version of SYMBOLIC FACTORABLE SUMT. The essential task in [3] was to associate relevant symbolic names with individual transformation codes which aided the identification of single variable functions in the model software.

Furthermore, this version permitted symbolic input for the problem variables and parameter names of up to four characters. In addition, it simplified the coding of the problems so that users could readily acquaint themselves with the symbolic names of various single variable functions, and could then code their problems without frequent reference to the manual. This was not the case in the earlier version, because the single variable functions had to be associated with the related transformation codes. Later, by exploiting the relative merits of factorable functions, deSilva and McCormick developed new routines to calculate a weighted cumulative sensitivity of the components of a solution vector with respect to problem parameters  $\sigma^T V_y x^*$  [Appendix 2, Equation (37)]. The underlying theory for the above developments is fully explained in [4], which is included in Appendix 2 for reference by the interested user.

This latter version has recently been further modified and refined by the authors to calculate sensitivity information for the individual components of the solution vector  $x^*$  with respect to the problem parameters  $y$  (i.e.,  $\frac{dx_i^*}{dy_j}$ ) and corresponding elasticities (i.e.,  $\frac{dx_i^*/x_i^*}{dy_j^*/y_j^*}$ ) for  $i=1, \dots, n$ ;  $j=1, \dots, k$ . In order to enhance its scope, a few more functions of a single variable have been added to the model. This latest version of the model is compiled at The George Washington University IBM/370, and can be accessed with the aid of job control cards, as depicted in Figures 2 or 3 of Section 4.

## 2. Factorable Functions

Most functions of several variables used in nonlinear optimization are complicated compositions of transformed sums and products of functions of a single variable. Once this observation is made formal, an outline of an automatic computer-oriented way of representing nonlinear programming problems becomes clear.



*Definition:* A function  $f(x_1, \dots, x_n)$  is a *factorable function* of several variables if it can be represented as the last in a finite sequence of functions  $f^j(x)$ , which are composed as follows:

$$f^j(x) = x_j, \quad \text{for } j=1, \dots, n.$$

For  $n < j$ ,  $f^j(x)$  is either of the form

$$f^k(x) + f^l(x), \quad k, l < j,$$

or of the form

$$f^k(x) \cdot f^l(x), \quad k, l < j,$$

or of the form

$$T[f^k(x)], \quad k < j,$$

where  $T(f)$  is a function of a single variable.

It should be emphasized that this is a *natural* way of looking at functions of several variables, since it corresponds to the way they are used when evaluated at a particular set of values of the vector  $(x_1, \dots, x_n)$ .

When a function is represented in factorable form, the computation of its gradient vector and Hessian matrix of second derivatives becomes very simple. Consider the general form for the gradient and Hessian matrix of a function which is either the sum of two functions, the product of two functions, or a transformation (function of one variable) of a function. For example,

$$\nabla[f(x)+g(x)] = \nabla f(x) + \nabla g(x)$$

$$\nabla[f(x) \cdot g(x)] = \nabla f(x) \cdot g(x) + \nabla g(x) \cdot f(x) \quad (1)$$

$$\nabla T[f(x)] = \nabla f(x) \cdot \dot{T}[f(x)], \quad (2)$$

where  $\dot{T}(f) = \partial T(f) / \partial f$ ; and

$$\begin{aligned}
\nabla^2[f(x)+g(x)] &= \nabla^2 f(x) + \nabla^2 g(x) \\
\nabla^2[f(x) \cdot g(x)] &= \nabla^2 f(x) \cdot g(x) + \nabla^2 g(x) \cdot f(x) \\
&\quad + \nabla f(x) \cdot \nabla^T g(x) + \nabla g(x) \cdot \nabla^T f(x) ,
\end{aligned} \tag{3}$$

$$\nabla^2[T[f(x)]] = \nabla^2 f(x) \cdot \dot{T}[f(x)] + \nabla f(x) \ddot{T}[f(x)] \nabla^T f(x) , \tag{4}$$

where  $\ddot{T}(f) = \partial^2 T(f) / \partial f^2$ .

Several points should be noted.

- (i) The computation of (1) involves  $f(x), g(x)$ , which would already have been computed in order to evaluate their product.
- (ii) The computation of (3) involves  $\nabla f(x), \nabla g(x)$ , which would already have been computed in order to evaluate (1).
- (iii) Equation (4) used  $\dot{T}[f(x)]$ , which was previously needed in (2).
- (iv) By induction, it follows that Hessian matrices of factorable functions are naturally given as sums of outer products of vectors, i.e.,

$$\sum u_i(x) \alpha_i(x) v_i^T(x) + v_i(x) \alpha_i(x) u_i^T(x) ,$$

where  $u_i(x), v_i(x)$  are  $n \times 1$  vectors,  $\alpha_i(x)$  are scalars, and the  $u_i(x), v_i(x)$  are available, having been required for the gradient computation.

For those optimization problems which are factorable, the factorable nonlinear programming problem is written in a more compact form (often referred to as canonical factorable form):

$$\begin{aligned}
&\text{minimize } X^N(x) \\
&\quad x \in E^n \\
&\text{subject to } L_i \leq X^1(x) \leq U_i ,
\end{aligned}$$

for  $i=1, \dots, N-1$  (possibly  $L_i = -\infty$  and/or  $U_i = +\infty$ ), where  $X^i(x) \equiv x_i$  (for  $i=1, \dots, n$ ), and the remainder are defined recursively as follows: given  $X^p(x)$ , for  $p=1, \dots, i-1$ , then for  $i=n+1, \dots, N$ ,

$$X^i(x) = \sum_{p=1}^{i-1} T_p^i[X^p(x)] + \sum_{p=1}^{i-1} \sum_{q=1}^p v_{q,p}^i[X^p(x)] \cdot U_{p,q}^i[X^q(x)], \quad (5)$$

where the  $T$ 's,  $U$ 's, and  $V$ 's are functions of a *single* variable. The functions  $X^i(x)$ ,  $i=1, \dots, N$  are usually referred to as concomitant variable functions and abbreviated as CVFs. Also,  $X^i(x)$ ,  $i=n+1, \dots, N$  are referred to as new CVFs.

An important point bears mentioning here. One of the many appealing features of coding and solving problems in factorable form is that they lend themselves in a natural way to the computation of the number of functional evaluations and arithmetic operations involved in the evaluation of problem functions, and in the first and second derivatives required for solution of nonlinear optimization problems. By doing so, a precise quantitative measure is provided for comparing algorithms devised for solving nonlinear optimization problems.

Following are two examples that demonstrate these ideas.

*Example i* [putting a function into factorable form].

Consider the following function:

$$W = \Gamma \cdot B(x_1 + x_2)(\ell_t x_4 + \ell_m x_5 + \ell_b x_6)[x_1 + (x_2^2 - x_3^2)^{1/2}]^{-1},$$

which corresponds to the objective function of the sample problem 3 provided in Section 4.3, where

$$\Gamma \cdot B = 3736.6$$

$$\ell_t = 495$$

$$\ell_m = 385$$

$$\ell_b = 315.$$

Table 1 shows how this function is put into simple factorable form, and depicts the corresponding stepwise reconstruction of the function.

*Example ii* [putting an optimization problem into canonical factorable form].

Consider the following optimization problem, which corresponds to the sample example 1 provided in Section 4.1:

$$\begin{aligned} & \underset{x_1, x_2}{\text{minimize}} \quad f = (x_1 - 10)^2 + (x_2 - 1)^2 \\ & \text{subject to} \quad g_1: \quad C_5 x_2 - 5 \sin .5x_1 \leq C_1 \\ & \quad \quad \quad g_2: \quad x_2 - C_6 \ln x_1 \geq C_2 \\ & \quad \quad \quad g_3: \quad x_1 x_2 \geq C_3 \\ & \quad \quad \quad g_4: \quad x_1 \geq C_4, \end{aligned}$$

where

$$C_1 = C_2 = 0, \quad C_3 = 3, \quad C_4 = 1.5, \quad C_5 = 1, \quad C_6 = -1.5$$

are the parameters of the problem. Putting the functions of this problem into simple factorable form goes as follows:

$$\begin{aligned} X^1(x) &= x_1 && \text{(functional form of } g_4) \\ X^2(x) &= x_2 \\ X^3(x) &= .5X^1(x) \\ X^4(x) &= \sin[X^3(x)] \\ X^5(x) &= -5X^4(x) \\ X^6(x) &= C_5 \cdot X^2(x) \\ X^7(x) &= X^6(x) + X^5(x) && \text{(functional form of } g_1) \\ X^8(x) &= \ln[X^1(x)] \\ X^9(x) &= -C_6 \cdot X^8(x) \\ X^{10}(x) &= X^2(x) + X^9(x) && \text{(functional form of } g_2) \end{aligned}$$



TABLE 1  
EXAMPLE OF FACTORABLE FUNCTION (EXAMPLE i)

Function in Sequence	Form of Composition	Portion of Original Function	Composition Type
$f^j(x)$	$x_j, j=1,6$		$\Gamma_j(x) = x_j$
$f^7(x)$	$f^1(x) + f^2(x)$	$x_1 + x_2$	$f + g$
$f^8(x)$	$\Gamma B \cdot f^7(x)$	$\Gamma B(x_1 + x_2)$	$T(f)$
$f^9(x)$	$\ell_t \cdot f^4(x)$	$\ell_t x_4$	$T(f)$
$f^{10}(x)$	$\ell_m \cdot f^5(x)$	$\ell_m x_5$	$T(f)$
$f^{11}(x)$	$\ell_b \cdot f^6(x)$	$\ell_b x_6$	$T(f)$
$f^{12}(x)$	$f^9(x) + f^{10}(x)$	$\ell_t x_4 + \ell_m x_5$	$f + g$
$f^{13}(x)$	$f^{12}(x) + f^{11}(x)$	$\ell_t x_4 + \ell_m x_5 + \ell_b x_6$	$f + g$
$f^{14}(x)$	$[f^2(x)]^2$	$x_2^2$	$T(f)$
$f^{15}(x)$	$-[f^3(x)]^2$	$-x_3^2$	$T(f)$
$f^{16}(x)$	$f^{14}(x) + f^{15}(x)$	$x_2^2 - x_3^2$	$f + g$
$f^{17}(x)$	$[f^{16}(x)]^{.5}$	$(x_2^2 - x_3^2)^{.5}$	$T(f)$
$f^{18}(x)$	$f^1(x) + f^{17}(x)$	$x_1 + (x_2^2 - x_3^2)^{.5}$	$f + g$
$f^{19}(x)$	$[f^{18}(x)]^{-1}$	$[x_1 + (x_2^2 - x_3^2)^{.5}]^{-1}$	$T(f)$
$f^{20}(x)$	$f^8(x) \cdot f^{13}(x)$	$\Gamma B(x_1 + x_2)(\ell_t x_4 + \ell_m x_5 + \ell_b x_6)$	$f \cdot g$
$f^{21}(x)$	$f^{20}(x) \cdot f^{19}(x)$	original function	$f \cdot g$



$$x^{11}(x) = x^1(x) \cdot x^2(x) \quad (\text{functional form of } g_3)$$

$$x^{12}(x) = x^1(x) - 10$$

$$x^{13}(x) = [x^{12}(x)]^2$$

$$x^{14}(x) = x^2(x) - 1$$

$$x^{15}(x) = [x^2(x) - 1]^2$$

$$x^{16}(x) = x^{13}(x) + x^{15}(x) \quad (\text{objective function } f)$$

Representing  $x^i(x)$  by  $x_i$  for notational simplicity, one canonical form of this problem will read as follows:

$$\begin{array}{ll} \text{minimize} & x_{16} \\ & x_1, \dots, x_{15} \end{array}$$

$$\text{where } x_1 = x_1 \quad (\text{functional form of } g_4)$$

$$x_2 = x_2$$

$$x_3 = .5x_1$$

$$x_4 = \sin x_3$$

$$x_5 = -5x_4$$

$$x_6 = c_5 x_2$$

$$x_7 = x_6 + x_5 \quad (\text{functional form of } g_1)$$

$$x_8 = \ln x_1$$

$$x_9 = -c_6 x_8$$

$$x_{10} = x_2 + x_9 \quad (\text{functional form of } g_2)$$

$$x_{11} = x_1 \cdot x_2 \quad (\text{functional form of } g_3)$$

$$x_{12} = x_1 - 10$$

$$x_{13} = (x_{12})^2$$

$$x_{14} = x_2 - 1$$

$$x_{15} = (x_{14})^2$$

$$x_{16} = x_{13} + x_{15} \quad (\text{objective function})$$

$$\text{subject to } x_7 \leq c_1 \quad (g_1)$$

$$\begin{aligned}
 x_{10} &\geq c_2 & (g_2) \\
 x_{11} &\geq c_3 & (g_3) \\
 x_1 &\geq c_4 & (g_4) .
 \end{aligned}$$

By exploiting the definition of the concomitant variable functions given in (5), many of the above steps may be combined and the problem may be reformulated (as is done in symbolic codings) more concisely as:

$$\begin{aligned}
 &\text{minimize } x^6 \\
 &\quad x \in E^6 \\
 &\quad \text{where } x^1 = x_1 && (\text{functional form of } g_4) \\
 &\quad \quad x^2 = x_2 \\
 &\quad \quad x^3 = c_5 x^2 - 5 \sin .5x^1 && (\text{functional form of } g_1) \\
 &\quad \quad x^4 = x^2 - c_6 \ln x^1 && (\text{functional form of } g_2) \\
 &\quad \quad x^5 = x^1 \cdot x^2 && (\text{functional form of } g_3) \\
 &\quad \quad x^6 = (x^1 - 10)^2 + (x^2 - 1)^2 && (\text{objective function}) \\
 &\text{subject to } x^3 \leq c_1 && (g_1) \\
 &\quad \quad x^4 \geq c_2 && (g_2) \\
 &\quad \quad x^5 \geq c_3 && (g_3) \\
 &\quad \quad x^1 \geq c_4 && (g_4) .
 \end{aligned}$$

### 3. Input Specifications

This section describes how the user must code and supply the required information about his problem to SYMBOLIC FACTORABLE SUMT. In the following, the type of information and its format, name, and description are discussed in the same order that must appear in a coded deck. A schematic depiction of a coded problem deck structure is provided in Figure 1 at the end of this section. For clarification, frequent reference to the code of sample problem 1 of Section 4.1 and to Figures 2 or 3 is advised for beginners as they proceed through the input specifications.

1 *JOB and JCL cards.* (See Figure 2, Section 4.1.4.)

2 *The SUMT parameter card.* The SUMT parameter card, which is read in by the subroutine MAIN, is described in Table 2.

TABLE 2  
THE SUMT PARAMETER CARD

Columns	Format	Name	Use
01-12	E12.0	EPSI ( $\epsilon$ )	Tolerance used to decide if an unconstrained minimum has been achieved for each subproblem (see Option 9)
13-24	E12.0	RHOIN ( $r_1$ )	Possible initial value of $r$ (often set at 1.0) (see Option 1)
25-36	E12.0	THETAO ( $\theta_0$ )	Tolerance used to decide if the solution to the NLP problem (A) has been approximated (see Option 5)
37-48	E12.0	RATIO ( $c$ )	Parameter ( $> 1$ ) used to compute consecutive values of $r$ ; $r_{i+1} = r_i/c$ (often set at 16.0)
49-60	E12.0	TMMAX	Maximum amount of time for solving problem (in seconds)
61-64	I4	M	Number (integer) of inequality constraints, $(M+MZ) \leq 200^*$
65-68	I4	N	Number (integer) of variables, $N \leq 100$
69-72	I4	MZ	Number (integer) of equality constraints

\*The limits on  $M+MZ$  and  $N$  are governed by the sizes of the arrays in SUMT.

3 *The first option card.* The first option card, which is read in by the subroutine MAIN, is described in Table 3.

TABLE 3

## THE FIRST OPTION CARD

Option	Column	Format	Name	Value	Meaning
1 (normally set to 3)	7	I	NT1	3	
2	14	I	NT2	2	
3 (normally set to 1)	21	I	NT3	1	Standard printout (this includes a call to OUTPUT after the solution of every subproblem). Also the estimates of the "Lagrange multipliers" and first- and second-order solution estimates are printed.
				2	For additional printout (includes standard printout and every intermediate point, gradient of P, and the vector S).
4 (normally set to 1)	28	I	NT4	1	Final convergence is determined on the basis of current solution to the subproblem.
				2	Final convergence is determined on the basis of the first order estimates. The first order estimate of the solution vector must be close to feasible. See below.
				3	Final convergence is determined on the basis of the second order estimates. The second order estimate of the solution vector must be close to feasible before the convergence check is made. If $\bar{x}$ is a solution estimate it is considered close to being feasible if $g_1(x) + \theta_0 \geq 0$ , $i=1,2,\dots,m$ , where $\theta_0$ is defined on the parameter card.



TABLE 3--continued

Option	Column	Format	Name	Value	Meaning
5 (normally set to 1)	35	I	NT5		The convergence criterion determining the NLP problem has been solved (only use = 1 when NT4 $\neq$ 1)
				1	Quit when $\frac{G - f[x(r_k)]}{G[x(r_k), \mu(r_k), \lambda(r_k)]} < \theta_0$
				2	Quit when $ r \sum_{j=1}^m \ln g_j[x(r_k)]  < \theta_0$
				3	Quit when first order estimate of $v_0$ $\frac{G[x(r_k), \mu(r_k), \lambda(r_k)]}{1} < \theta_0$
6 (normally set to 1)	42	I	NT6	1	After final convergence the program reads in new data and solves the next problem
				2	After final convergence has been determined a call to PUNCH is made before proceeding on to the next problem
7 (normally set to 1)	49	I	NT7		First move after a minimum to a subproblem is achieved
				1	No extrapolation
				2	Extrapolate through last two minima
8 (normally set to 0)	56	I	NT8	3	Extrapolate through last three minima
				0	No sensitivity or elasticity calculation
				2	Conduct sensitivity and elasticity calculation after each subproblem convergence

TABLE 3--continued

Option	Column	Format	Name	Value	Meaning
9	63	I	NT9		Subproblem convergence criterion, or when to stop minimizing P function for fixed value of r (see parameter card)
				1	Quit when $\left  \nabla_x P^T(x^i, r) \left[ \frac{\partial^2 P(x, r)}{\partial x_i \partial x_j} \right]^{-1} \right $ $ \nabla_x P(x^i, r)  < \epsilon$
				2	Quit when $\left  \nabla_x P^T(x^i, r) \left[ \frac{\partial^2 P(x, r)}{\partial x_i \partial x_j} \right]^{-1} \right $ $ \nabla_x P(x^i, r)  < \frac{P(x^{i-1}) - P(x^i)}{5}$
10	70	I	NT10	3	Quit when $ \nabla_x P(x^i, r)  < \epsilon$
				1	At least one nonlinear constraint
				2	Linear constraints
				3	Linear constraints and linear objective function (i.e., a linear programming problem)

4 *The title card.* This card, which is read in by the subroutine SYMPUT, contains all pertinent information such as problem title and reference, user's name, date, etc., that the user wishes to appear on the output. This information must be coded in a single card anywhere in columns 1-80, and is read in with the alphanumeric format of 20 A4. A blank card must be supplied if this information is not coded.

5 *The SYMPUT parameter card.* The SYMPUT parameter card, which is read in by the subroutine SYMPUT, is described in Table 4.

TABLE 4  
THE SYMPUT PARAMETER CARD

Columns	Format	Name	Description
01-05	I <sub>5</sub>	N	Number of problem variables
06-10	I <sub>5</sub>	NN	Number of concomitant variable functions, CVFs (i.e., number of variables in factorable form)
11-15	I <sub>5</sub>	MMM	Number of inequality constraints
16-20	I <sub>5</sub>	MZ	Number of equality constraints
21-25	I <sub>5</sub>	ITEST	<ul style="list-style-type: none"> <li>• If ITEST=0, at the initial feasible point of the problem the values and first and second derivatives of new CVFs will be evaluated and printed for debugging purposes, without solving the problem</li> <li>• If ITEST=1, the above information will be suppressed and the problem will be solved</li> </ul>
26-30	I <sub>5</sub>	IPR	<ul style="list-style-type: none"> <li>• If IPR=1, after each subproblem convergence the values of all CVFs at the subproblem solution point will be printed.</li> <li>• If IPR=0 (<math>\neq 1</math>), the values of the CVFs will be suppressed after each subproblem convergence</li> </ul>
31-35	I <sub>5</sub>	ITES	<ul style="list-style-type: none"> <li>• If ITES=2, after each subproblem convergence the following information will be printed for debugging purposes: <ul style="list-style-type: none"> <li>(i) Entries of vector <math>\sigma^{(1)}</math></li> <li>(ii) Entries of vector <math>\alpha^{(2)}</math></li> <li>(iii) Values of all CVFs</li> <li>(iv) Partial derivatives of new CVFs with respect to problem variables and parameters</li> <li>(v) Entries of vector <math>\text{DELA}^{(3)}</math></li> </ul> </li> <li>• If ITES=0 (<math>\neq 2</math>), the above information will be suppressed after each subproblem convergence</li> </ul>

TABLE 4--continued

Columns	Format	Name	Description
36-40	I <sub>5</sub>	NSWIT	<ul style="list-style-type: none"> <li>• If NSWIT=1, the modified listing of the coded problem and related tables set up by subroutine SYMPUT will be suppressed in the output</li> <li>• If NSWIT=0 (<math>\neq 1</math>), the above listing will not be suppressed</li> </ul>
41-45	I <sub>5</sub>	IID	Number of problem parameters (i.e., P and D cards)

(1) Vector  $\sigma$  is an  $n$ -dimensional unit vector  $e_i$ ,  $i=1, \dots, n$  ( $n$  is the number of original problem variables).

(2) See Equation (38) in Appendix 2.

(3) Negative values of partial derivatives of subproblem solution vector components with respect to the problem parameters.

6 *The variable definition cards.* These cards are read in by the subroutine SYMPUT. For each variable of the problem, the user must supply one card with the descriptions as shown in Table 5.

TABLE 5

THE VARIABLE DEFINITION CARDS

Columns	Format	Name	Description
02	A1	ATYPE	Letter "v" must be coded for ATYPE to indicate this card corresponds to a variable card
04-07	A <sub>4</sub>	ANAME	Variable name to be specified by the user (right justified)
16-28	F12.5	ARL	Initial value of variable (starting point)

It is obvious that the number of variable definition cards should match the number of the (original) variables of the problem being coded.



7 *The problem parameter cards (P and D cards).* These cards are read in by the subroutine SYMPUT. For each parameter of the problem, the user must supply one card with a description as shown in Table 6.

TABLE 6  
THE PROBLEM PARAMETER CARDS

Columns	Format	Name	Description
02	A <sub>1</sub>	ATYPE	<ul style="list-style-type: none"> <li>• Letter "D" must be coded for ATYPE to indicate this card corresponds to a parameter card</li> <li>• Use of letter "P" instead of letter "D", in conjunction with proper option in the SUMT parameter card, will invoke the sensitivity routines to calculate the sensitivity and elasticities of the solution point with respect to the indicated parameter after each subproblem convergence</li> </ul>
04-07	A <sub>4</sub>	ANAME	Parameter name to be specified by the user (right justified)
16-28	F12.5	ARG	Numerical value of the parameter

It is obvious that the number of the parameter cards must match the number of problem parameters. Also, parameter cards with "P", on which sensitivity calculation is desired, *must* appear before parameter cards with letter "D".

8 *The separable and quadratic cards.* These cards are read in by the subroutine SYMPUT. The number and order of appearance of separable and quadratic cards will depend on the problem functions being coded and their canonical formulation by the user. Each separable or quadratic card in essence provides the model with information regarding a single variable function. With this information, the model can systematically synthesize the new concomitant variable functions and subsequently can reconstruct the problem functions and their derivatives for the purposes of various calculations.

While coding these cards, the user must be aware of the following facts:

- Separable cards may appear sequentially in any numbers. As long as they are associated with the same new CVF (CVF in columns 04-07 of Table 7), the single variable functions in each card will be *summed up* recursively by the model to construct the new CVF.
- Quadratic cards must appear as pairs, with any number of pairs permitted. As long as these cards are associated with the same new CVF, the *product* of the related single variable functions in each consecutive pair will be *summed up* recursively (if there are more than one pair) to reconstruct the new CVF.
- When there are separable *and* quadratic cards associated with the same new CVF, the results of the aforementioned calculations will be *summed up* to reconstruct the subject CVF.

Separable and quadratic cards are described in Table 7. For further clarification, the reader is referred to the formulation and coding of the sample problems provided in Section 4.

9 *The bound cards.* These cards are read in by the subroutine SYMPUT. There are two types of bound cards, upper bound and lower bound. It is natural that these cards must appear after the concomitant variable functions that are being bounded have been defined. Cards relating to equality constraints *must* follow the cards defining the inequality constraints. Table 8 describes bound cards.

10 *The flag card.* The user must indicate the end of his problem's symbolic factorable code by furnishing a flag card. This card, which is read in by the subroutine SYMPUT, must contain the number 5 in its second column.

11 *The tolerance card.* This card is read in by the subroutine MAIN, and is described in Table 9.

TABLE 7  
THE SEPARABLE AND QUADRATIC CARDS

Columns	Format	Name	Description
02	A1	ATYPE	Letter "S" must be coded for ATYPE if the card is a separable card. Letter "Q" must be coded for ATYPE if the card is a quadratic card
04-07	A4	ANAME	Name of the new CVF must be coded for ANAME (right justified)
09-12	A4	ARG	Name of a previously defined CVF, in terms of which the current new CVF is being defined, must be coded for ARG
14-16	A4	AFUN	Three-letter symbolic name assigned to the type of functional relationship between ANAME and ARG (see Table 12) must be coded for AFUN
17-28	F12.5 or 3A4	AC1†	<ul style="list-style-type: none"> <li>• The value of C1 in AFUN (see Table 12) must be coded for AC1</li> <li>• If C1 corresponds to a previously defined problem parameter, then the parameter name must be coded for AC1 (right justified)*</li> </ul>
29-40	F12.5 or 3A4	AC2†	<ul style="list-style-type: none"> <li>• The value of C2 in AFUN (see Table 12) must be coded for AC2</li> <li>• If C2 corresponds to a previously defined problem parameter, then the parameter name must be coded for AC2 (right justified)*</li> </ul>

†Reading of AC1 and/or AC2 with the proper F or A format in the model is made possible by utilizing the subroutine REREAD, which is a built-in subroutine at The George Washington University FORTRAN library (see Section 5).

\*Parameters with alphanumeric value *cannot* take negative signs throughout the code (i.e., -ABC does not imply the negative of numeric value of parameter ABC).

TABLE 8  
THE BOUND CARDS

Columns	Format	Name	Description
02	A1	ATYPE	Letter "U" must be coded for ATYPE if an upper bound is being imposed, letter "L" for a lower bound. An equality constraint uses either letter.
04-07	A4	ANAME	The name of the first of a sequence of CVFs to which the bound applies must be coded for ANAME (by sequence it is meant the order in which the CVFs appear in the code)
09-12	A4	ARG	The name of the last CVF in the above (uninterrupted) sequence to which the bound applies must be coded for ARG
17-28	F12.5 or 3A4	AC1†	<ul style="list-style-type: none"> <li>• The value of the bound for the above CVF must be coded for AC1</li> <li>• If the bound corresponds to a previously defined parameter, then the parameter name must be coded for AC1 (right justified)</li> </ul>

†See the footnote of Table 7.

TABLE 9  
THE TOLERANCE CARD

Columns	Format	Name	Description
01-12	E12.6	XEP1	Ignored by the program
13-24	E12.6	XEP2	When minimizing the P-function for a given value of $r$ ( $RH\emptyset$ ) the value of P must decrease by an amount exceeding XEP2 for each iteration after the first. If it does not, then the code prints out the message "apparently roundoff errors prevent a more accurate determination of the minimum of this subproblem," and it is assumed that a minimum has been found. (Usually we set XEP2 equal to 0.)



12 *The second option card.* This card is read in by the program MAIN. The description of this card is given in Table 10.

TABLE 10  
THE SECOND OPTION CARD

Option	Column	Value	Meaning
1	7	1	Ignored in this program
2	14	1	The method for minimizing the unconstrained penalty function is to be the generalized Newton-Raphson method as modified to handle indefinite Hessian matrices. This method requires function values, first and second derivatives
		2	Same as 1, except that when an "orthogonal move" is made because of an indefinite Hessian matrix, $-VP$ is added to the orthogonal move vector
		3	Steepest descent is used to minimize the P-function
		4	The method for minimizing the unconstrained penalty function is the rank one method as reported in the Fiacco-McCormick book [10]. This requires function values and first derivatives

13 *End card.* This is the usual end card which must follow any coded problem. It contains "/" at the first two columns.

A schematic depiction of a problem deck structure is shown in Figure 1.

#### 4. Coded Examples and Output Illustrations

For illustration purposes the coded input and the resulting output for the sample problem 1 discussed in Section 2 is reviewed in sufficient detail. Although the functional relationships involved in this problem are too simple and the problem itself is too small to illustrate the

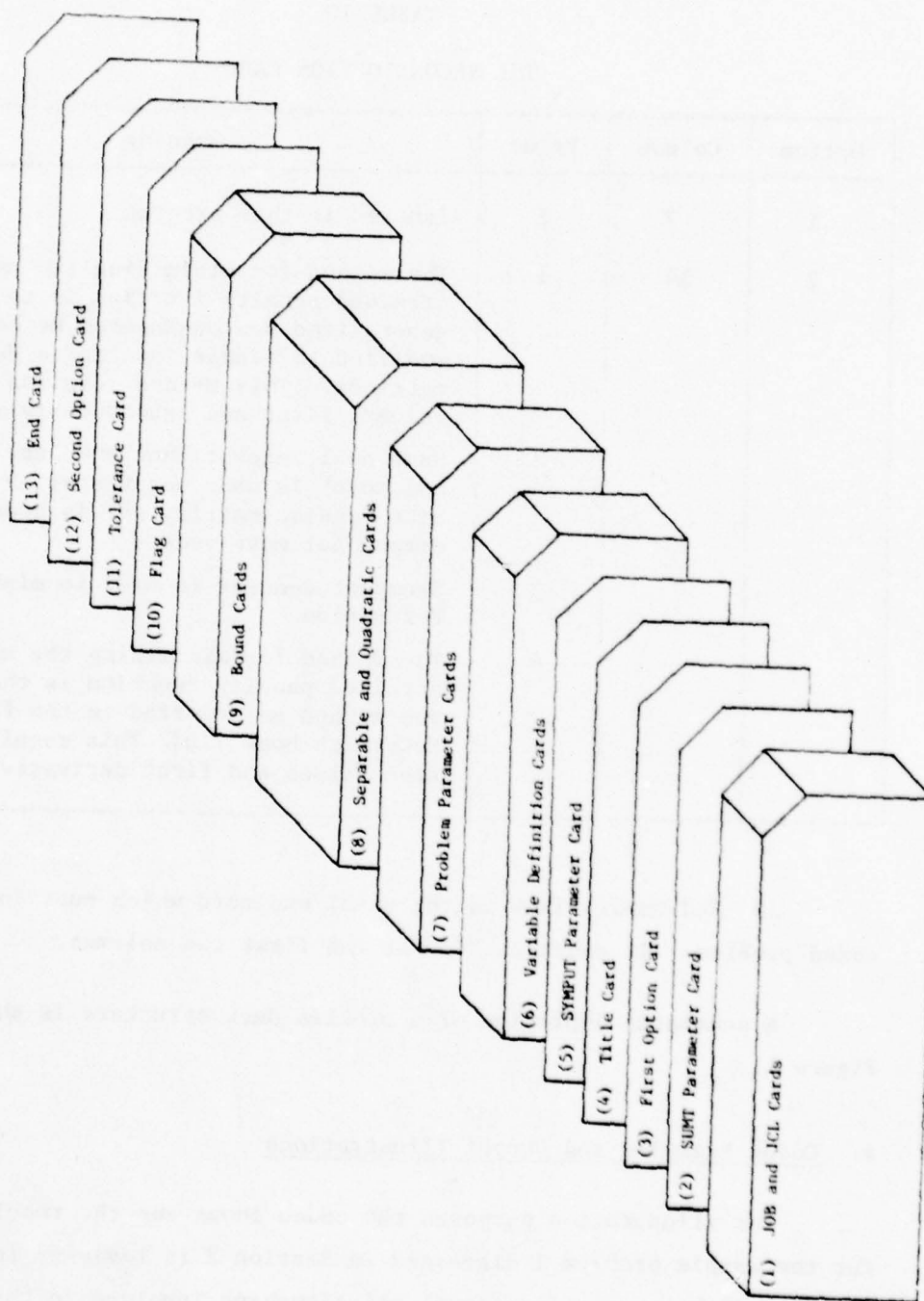


Figure 1.--Data deck structure.

merits of SYMBOLIC FACTORABLE SUMT, we have purposely chosen it so that users can readily follow and become familiar with the model with the least possible effort. For further elucidation the input/output for two more sample problems--Hopkin's problem from Reference [2] and the bulk-head design problem from Reference [5]--are included following the sample problem 1.

#### 4.1 Sample problem 1 (a test problem).

##### 4.1.1. Problem definition (restated for ease of reference):

$$\begin{aligned}
 &\underset{x_1, x_2}{\text{minimize}} && (x_1 - 10)^2 + (x_2 - 1)^2 \\
 &\text{subject to} && C_5 x_2 - 5 \sin .5x_1 \leq C_1 \\
 & && x_2 - C_6 \ln x_1 \geq C_2 \\
 & && x_1 x_2 \geq C_3 \\
 & && x_1 \geq C_4 ,
 \end{aligned}$$

where  $C_1$  through  $C_6$  are the parameters of the problem, and their respective values are

$$C_1 = C_2 = 0 , \quad C_3 = 3 , \quad C_4 = 1.5 , \quad C_5 = 1 , \quad C_6 = -1.5 .$$

Moreover, let us suppose that we intend to calculate the sensitivity and elasticity of the solution vector with respect to the parameters  $C_5$  and  $C_6$ .

4.1.2. Symbolic factorable form. In SYMBOLIC FACTORABLE SUMT, use of canned routines to evaluate functions of one variable in a single iteration enables us to code the problem in the following concise form:

minimize $X^6$	<i>symbolic</i>
$X \in E^6$	<i>factorable</i>
where $X^1 = x_1$	<u>step</u>
	(a)
$X^2 = x_2$	(b)
$X^3 = C_5 X^2 - 5 \sin .5 X^1$	(c)
$X^4 = X^2 - C_6 \ln X^1$	(d)
$X^5 = X^1 \cdot X^2$	(e)
$X^6 = (X^1 - 10)^2 + (X^2 - 1)^2$	(f)
subject to $X^3 \leq C_1$	(g)
$X^4 \geq C_2$	(h)
$X^5 \geq C_3$	(i)
$X^1 \geq C_4$	(j)

4.1.3. The keypunch sheet code. The keypunch sheets used to code the above problem in symbolic factorable language is shown in Figure 2. Table 11 shows the correspondence between the steps cited in 4.1.2 and the related symbolic code. Figure 3 depicts an annotated computer listing of the symbolic code for problem 1. The categories of the code are numbered and specified in the left margin of this figure in the same sequence as was illustrated in Section 3.

The symbolic names used in the code are directly taken from Table 12 at the end of the present section.

4.1.4. Output annotations. The computer output for the sample example 1 is depicted in Appendix 1, page 50. Various segments of this output are numbered in circles. Corresponding explanations are as follows:

- 1 Information contained in the title card
- 2 Printout of SYMPUT parameter card (the reader may skip reading steps 3-8).



```

11AAAA JOB (5660,R),ABG
11 EXEC F0RG6,DSN='0R5660.ABG',PR06=MAIN
11SYSLIB DD
11      DD
11      DD DSN=GNU.F0RTL1B,DISP=SHR
11G0.FT05F001 DD *

```

EXAMPLE-1 : SOLUTION AND SENSITIVITY ANALYSIS															T-402	
2	6	4	0	1	0	0	0	6								
V	X1								2.0							01
V	X2								2.0							02
P	C5								1.0							03
P	C6								-1.5							04
D	C1								0.0							05
D	C2								0.0							06
D	C3								3.0							07
D	C4								1.5							08
S	X3	X1 SIN							0.5							09
S	X3	X2 LIN							C5							10
S	X4	X2 LIN							1.0							11
																12
																13
																14
																15

Figure 2.--Data cards for sample problem 1.

Variable	Value	Variable	Value
S	1.0	X4	1.0
Q	1.0	X5	1.0
Q	1.0	X5	1.0
S	1.0	X6	1.0
S	1.0	X6	1.0
U	1.0	X3	1.0
L	1.0	X4	1.0
L	1.0	X5	1.0
L	1.0	X1	1.0
S	1.0	X1	1.0

Figure 2.--continued.



TABLE 11

THE CORRESPONDENCE BETWEEN FUNCTIONS OF SAMPLE PROBLEM 1 IN  
SYMBOLIC FACTORABLE FORM AND ITS SYMBOLIC FACTORABLE CODE

Symbolic Factorable Step	Statement Serial Number in Key punch Sheet (Figure 2)
(a)	05
(b)	06
(c)	13 and 14
(d)	15 and 16
(e)	17 and 18
(f)	19 and 20
(g)	21
(h)	22
(i)	23
(j)	24

Segments 3-8 pertain to the categorization of the coded symbolic input by the subroutine SYMPUT which is necessary for synthesis of the original problem from the symbolic code. In order to make these print-outs self-explanatory, the reader should notice the following points, which are related to the programming logic of the subroutine SYMPUT:

- (i) Separable terms are labeled with a type number "3" and quadratic terms with a type number "4."
- (ii) CVFs are sequentially labeled from 1 to NN (total number of CVFs) in the order of their appearance in the code.
- (iii) All the problem parameters and constant coefficients are sequentially labeled by natural sequence of numbers in the order of their appearance in the code and stored in array CCl. Constants having the same numerical value are labeled with the same number. Moreover, in this array the problem parameters are separated from problem constant coefficients with numbers zero and one.



- (iv) Constraints of the problem are sequentially labeled by natural sequence of numbers in the order of their appearance in the code.
- (v) Each symbolic function, as discussed in Section 1, is associated with a unique number called a transformation code, which serves as a link to juxtapose the symbolic function name with its mathematical code.

With these points in mind we proceed to review briefly segments 3-8 of the output.

3 Modified printout of the symbolic code of the problem (excluding the bounds). For explanation of columns 2, (5,7), and 8, see comments (i), (ii), and (v), respectively, above.

4 Manipulated printout of problem parameters and constants [see comment (iii)]

5 Bounds on problem constraints. For explanation see comments (iv) and (iii), respectively.

6 Number of times that a concomitant variable function (CVF) is defined in terms of a separable one-variable function or quadratic function (cumulative). For example, concomitant variable function  $X^4$  is defined as follows:

$$X^4 = X^2 - C_2 \ln X^1,$$

which has  $5-3 = 2$  separable terms and  $1-1 = 0$  quadratic terms.

7 Table of separable terms formed by subroutine SYMPUT (self-explanatory).

8 Table of quadratic terms set up by subroutine SYMPUT (self-explanatory).

9 Printout of parameter card.

10 Printout of first option card.

11 Printout of tolerance card.

- 12 Printout of second option card.
- 13 Printout of initial vector and corresponding problem function values.
- 14 Printout of feasible starting vector and corresponding problem function values (initial vector need not be feasible).
- 15 Objective function value, solution vector, constraint values, and Lagrange multipliers evaluated at solution point of subproblem 1 (at  $r = 100$ ) . The following information is also evaluated at this solution point:
- DOTT : inverse of Hessian matrix of penalty function  
pre- and post-multiplied by penalty function  
gradient
  - MAGNITUDE: norm of penalty function gradient
  - P : value of the penalty function
  - G : dual value
  - RSIGMA : residual term related to inequality constraints  
in penalty function
  - H : residual term related to equality constraint in  
penalty function.
- 16 Printout of the problem parameters on which sensitivity information is desired.
- 17 Zeroth order sensitivity of the first component of the solution vector of subproblem 1 (i.e.,  $x_1$ ) and corresponding elasticities with respect to the problem parameters (repeated for all components of solution vector).
- 18 Solution to the second subproblem (at  $r = 10$ ) and first order estimates of solution values.
- 19 First order sensitivity of the first component of solution vector of subproblem 2 and corresponding elasticities with respect to the problem parameters.
- 20 Solution to the third subproblem (at  $r = 1$ ) and first and second order estimates of solution values.
- 21 Solution to the fourth subproblem (at  $r = .1$ ) .

- 22 Solution to the fifth subproblem (at  $r = .01$ ) .
- 23 Solution to the sixth subproblem (at  $r = .001$ ) .
- 24 Solution to the seventh subproblem (at  $r = .0001$ ) .
- 25 Solution to the eighth (final) subproblem (at  $r = .00001$ ) .

Illustration of a few useful optional outputs seems beneficial at this stage.

- ITEST=0 rather than 1 (in SYMPUT parameter card) would suppress output segments numbered from 15 onwards, and would instead yield output segments A, B, C, and D, as labeled on page 64 of Appendix 1, and which have the following descriptions (all evaluated at the initial feasible point):
  - A : the values of CVFs
  - B : partial derivatives of the new CVFs with respect to the problem variables
  - C : diagonal entries of the Hessian matrix of the new CVFs
  - D :
- NSWIT=1 (in SYMPUT parameter card) would suppress the output segments 3-8.
- ITES=2 (in SYMPUT parameter card) would yield output segments E, F, G, H, I, J, and K after solution of each subproblem, which are defined as follows (all evaluated at the corresponding subproblem solution point):
  - E : the entries of vector  $\sigma$  (see Table 4)
  - F : the entries of vector  $\alpha$  [see Equation (38), Appendix 2]
  - G : the values of CVFs
  - H : partials of the new CVFs with respect to the original problem variables premultiplied by vector  $\alpha$
  - I : the joint partial derivatives of the new CVFs with respect to the original problem variables and sensitivity parameters premultiplied by vector  $\alpha$
  - J : the partials of the new CVFs with respect to the sensitivity parameters
  - K : the negative values of the zeroth order sensitivity of the original problem variables with respect to the sensitivity parameters.

- NT8=1 (in first option card) would suppress the sensitivity and elasticity analysis calculations.

#### 4.2 Sample problem 2 (Hopkins problem).

This problem has been solved in Reference [2] using FACTORABLE SUMT (earlier version of the SYMBOLIC FACTORABLE SUMT). Here it is chosen as the second sample problem for solution and sensitivity analysis by the SYMBOLIC FACTORABLE SUMT. For brevity, unlike the first sample problem, illustration of this sample problem is limited to the algebraic definition of the problem, computer listing of its code, and selected pages of the computer solution output.

##### 4.2.1. Problem definition.

$$\text{minimize } \left( -100x_1^{.5} - 2x_2 - 25x_3^{.5} - 50x_4^{.5} - 10x_5 + 100(x_6-3)^2 \right. \\ \left. - 30x_1^{.5}x_7^{.5} + 100x_8 - 10x_1/x_3 - 20x_1/x_4 \right)$$

$$\text{subject to } g_1 : A_1x_1 + A_2x_2 + A_3x_3 + A_4x_4 + A_5x_3x_6 + A_6x_4x_6 \\ - A_7x_4x_5x_6 + A_8x_1x_7 - A_9x_1x_8 \leq 25,000$$

$$g_2 : 12.25x_1 + x_3x_6 + .7x_4x_6 - .2x_4x_5x_6 \geq 34,000$$

$$g_3 : x_5 \leq .5$$

$$g_4 : x_8 \leq .3$$

$$g_5-g_{14} : x_j \geq 0; j=1,8,$$

where the coefficients of the first constraint, i.e.,  $A_1-A_9$ , are chosen as problem parameters on which sensitivity and elasticity analysis is performed. The values of these parameters are:

$$A_1 = 24, \quad A_2 = 12, \quad A_3 = .6, \quad A_4 = .4, \quad A_5 = .1 \\ A_6 = .15, \quad A_7 = -.15, \quad A_8 = 11, \quad \text{and} \quad A_9 = 24.$$



TABLE 12  
LIST OF SINGLE VARIABLE FUNCTIONS USED BY SYMBOLIC FACTORABLE SUMT

Symbolic Name	Description	Algebraic Form	Default Options
CØN	Constant	C1	None
IDN	Identity	Z	None
LIN	Linear	C2 + C1*Z	C2 = 0
EXP	Exponential	C2* EXP(C1*Z)	C2 = 1, C1 = 1
EXC	--	C2* (C1) <sup>Z</sup>	C2 = 1
PØW	Power	C2* (Z) <sup>C1</sup>	C2 = 1
PØL	Polynomial	C2*(Z) <sup>C1</sup> (C1 integer)	C2 = 1
LNG	Natural Log	C2* LN(Z)	C2 = 1 [no C1]
LØG	Other Log	C2* LØG <sub>C1</sub> (Z)	C2 = 1
SIN	Sine	C2* SIN(C1*Z)	C2 = 1, C1 = 1 <sup>+</sup>
CØS	Cosine	C2* CØS(C1*Z)	C2 = 1, C1 = 1 <sup>+</sup>
TAN	Tangent	C2* TAN(C1*Z)	C2 = 1, C1 = 1 <sup>+</sup>
GAM	Gamma Function	C2* Γ(C <sub>1</sub> *Z)	C2 = 1
NØR	Cumulative Normal	C2* $\frac{1}{\sqrt{2\pi}} \int_{-\infty}^Z \text{EXP}(-\frac{t^2}{2}) dt$	C2 = 1
HEV	Heavisides	C2* H(Z-C1)	C2 = 1

TABLE 12--continued

Symbolic Name	Description	Algebraic Form	Default Options
MIN	Minimum	$C2 * \text{MIN}[Z, C1]$	$C2 = 1$
MAX	Maximum	$C2 * \text{MAX}[Z, C1]$	$C2 = 1$
ARS	Arc Sine	$C2 * \text{ARCSIN}[C1 * Z]$	$C2 = 1$
ART	Arc Tangent	$C2 * \text{ARCTAN}[C1 * Z]$	$C2 = 1$
QDR	Quadratic	$[C1 * Z - C2]^2$	None
ESQ	Exponential Square	$C2 * \text{EXP}[C1 * Z^2]$	$C2 = 1$
ENT	Entropy Function	$C1 * Z [C2 + \text{LN}(C1 * Z)]$	None
LGS	Logistic Curve	$[1 + C2 * \text{EXP}(-C1 * Z)]^{-1}$	$C2 = 1$
EXM	Decay Function	$C2 * [1 - \text{EXP}(-C1 * Z)]$	$C2 = 1$
LLN	Log of Linear Function	$\text{LN}(-C1 * Z + C2)$	None
DIV	Division	$C1 / Z + C2$	$C1 = 1$

†Default option  $C1 = 1$  can be used only if the option  $C2 = 1$  is also being exploited, i.e.,  $C2$  is not coded.

‡Heaviside function is a step function defined as follows:

$$C2 * H(Z - C1) = \begin{cases} 0, & \text{if } Z < C1 \\ C2, & \text{if } Z \geq C1 \end{cases}$$

4.2.2. Computer listing of the symbolic code. (See Figure 4.)

4.2.3. Selected pages from computer output. (See Appendix 1, page 69.)

#### 4.3 Sample problem 3 (corrugated transverse bulkhead design).

This problem, which has been taken from Reference [5], involves more complicated functional relationships than the two previous sample problems. It has been solved by many authors [6], [5], [7], [8]. A comparison of the time, cost, and effort spent to code and solve this problem in [5] and [8]--where the SUMT-Version 4 model was used--with those involved in the present study, shows the enormous advantage of the SYMBOLIC FACTORABLE SUMT model over the SUMT-Version 4 model. Table 13 accounts for this claim.

TABLE 13

SUMT-VERSION 4 VERSUS SYMBOLIC FACTORABLE SUMT ON SOLUTION OF OPTIMAL CORRUGATED TRANSVERSE BULKHEAD DESIGN PROBLEM

Comparison Criteria	SUMT-Version 4 [8]	SYMBOLIC SUMT [present study]
Number of statements used to code the problem	300 <sup>+</sup>	100
Approximate time spent formulating and coding the problem, hrs	25 <sup>+</sup>	4
CPU time spent solving the problem/run, seconds	61.63	27.64
Cost incurred in solving the problem/run, dollars	5.08 <sup>+</sup>	2.73

A comparison of the entries in the above table reveals how SYMBOLIC FACTORABLE SUMT facilitates the coding of problems and brings about savings in time and costs involved in the solution of nonlinear problems. As mentioned in Section 1, the primary reason for such reductions, especially in the coding of the problem, stems from the fact that SUMT-Version

```
//AAAAAA JOB (5660,R,3,3),ABG
// EXEC FORG6,DSN='OR5660.ABG',PROG=MAIN
//SYSLIB DD
//      DD
//      DD DSN=GWU.FORTLIB,DISP=SHR
//GO.FT05F001 DD *
```

```
      0.00001      100.0      .00001      10.0      300.  12   8   0
      3          2          2          1          1          1          1          2          1          1
EXAMPLE-2 : SOLUTION AND SENSITIVITY ANALYSIS OF HOPKINS PROBLEM
      8   14   12   0   1   0   0   0   0   9
```

```
V  X1      180.00
V  X2      740.00
V  X3      2680.00
V  X4      14650.00
V  X5        .40
V  X6        2.30
V  X7        .60
V  X8        .20
P  A1      24.00
P  A2      12.00
P  A3       0.600
P  A4       0.40
P  A5       0.10
P  A6       0.15
P  A7      -0.15
P  A8       11.00
P  A9      -24.00
Q  X9      X4 IDN
Q  X9      X6 IDN
Q  X10     X3 IDN
Q  X10     X6 IDN
S  X11     X1 LIN      A1
S  X11     X2 LIN      A2
S  X11     X3 LIN      A3
S  X11     X4 LIN      A4
S  X11     X10 LIN     A5
S  X11     X9 LIN      A6
Q  X11     X9 LIN      A7
Q  X11     X5 IDN
Q  X11     X1 LIN      A8
Q  X11     X7 IDN
Q  X11     X1 LIN      A9
Q  X11     X8 IDN
S  X12     X1 LIN      12.25
S  X12     X10 IDN
S  X12     X9 LIN       0.70
Q  X12     X9 LIN      -0.20
Q  X12     X5 IDN
```

Figure 4.--Computer listing of sample problem 2.



S	X13	X6	QDR	1.00	3.
S	X14	X1	POW	0.50	-100.
S	X14	X2	LIN	-2.00	
S	X14	X3	POW	0.50	-25.
S	X14	X4	POW	0.50	-50.
S	X14	X5	LIN	-10.00	
S	X14	X13	LIN	100.00	
Q	X14	X1	POW	0.50	-30.
Q	X14	X7	POW	0.50	
S	X14	X8	LIN	100.00	
Q	X14	X1	LIN	-10.00	
Q	X14	X3	POW	-1.	
Q	X14	X1	LIN	-20.00	
Q	X14	X4	POW	-1.	
U	X11	X11		25000.00	
L	X12	X12		34000.00	
U	X5	X5		.50	
U	X8	X8		.30	
L	X1	X8		0.00	
5					
	0.001		.001		
//	1	1			

Figure 4.--continued

4 requires explicit derivations and coding of the problem function gradients and Hessians (user's subroutine), while in SYMBOLIC FACTORABLE SUMT these requirements are internally calculated and so are relaxed. The derivation and coding of these entries is sometimes extremely tedious and time consuming, as is the case for the present problem, and is prone to mathematical errors whose debugging is another factor in the user's time consumption. It was this very fact that motivated the development of SYMBOLIC FACTORABLE SUMT, and the present example clearly demonstrates that this goal has been very well achieved.

It is not the intention of the present work, or within its scope, to discuss the engineering aspects of this optimal design problem and the significance of its constraints. Readers interested in such aspects should refer to [5]. Instead, similar to the sample problem 2, its discussion is limited to the algebraic definition of the problem, a computer listing of its code, and selected pages from computer solution output.

4.3.1. Problem definition. Following is the algebraic definition of the problem, borrowed from [5]. For sensitivity analysis considerations, the coefficients of  $K_2$  and  $t^{\min}$  in constraints 8-16 have been normalized to unity in the code of the problem (i.e., constraints 8, 11, and 14 have been scaled by  $-1$ , and constraints 9, 10, 11, 12, 13, 15, and 16 by  $-\frac{1}{10}$ ). The problem is

$$\begin{aligned}
 &\text{minimize } W = \Gamma B(x_1 + x_2)(\ell_t x_4 + \ell_m x_5 + \ell_b x_6)[x_1 + (x_2^2 - x_3^2)^{\frac{1}{2}}]^{-1} \\
 &\text{subject to} \quad \begin{array}{rcl} & & \text{constraint} \\ & & \text{number} \\ & x_2 - x_3 \geq 0 & (1) \\ & \frac{1}{6} x_2 x_3 x_4 + \frac{e}{2} x_1 x_3 x_4 - K_1 h_t \ell_t^2 [x_1 + (x_2^2 - x_3^2)^{\frac{1}{2}}] \geq 0 & (2) \\ & \frac{1}{6} x_2 x_3 x_5 + \frac{e}{2} x_1 x_3 x_5 - K_1 h_m \ell_m^2 [x_1 + (x_2^2 - x_3^2)^{\frac{1}{2}}] \geq 0 & (3) \\ & \frac{1}{6} x_2 x_3 x_6 + \frac{e}{2} x_1 x_3 x_6 - K_1 h_b \ell_b^2 [x_1 + (x_2^2 - x_3^2)^{\frac{1}{2}}] \geq 0 & (4) \end{array}
 \end{aligned}$$

$$\frac{1}{12} x_2 x_3^2 x_4 + \frac{e}{4} x_1 x_3^2 x_4 - 2.2(K_1 h_t \ell_t^2)^{4/3} [x_1 + (x_2 - x_3)^{1/2}]^{4/3} \geq 0, \quad (5)$$

$$\frac{1}{12} x_2 x_3^2 x_5 + \frac{e}{4} x_1 x_3^2 x_5 - 2.2(K_1 h_m \ell_m^2)^{4/3} [x_1 + (x_2 - x_3)^{1/2}]^{4/3} \geq 0, \quad (6)$$

$$\frac{1}{12} x_2 x_3^2 x_6 + \frac{e}{4} x_1 x_3^2 x_5 - 2.2(K_1 h_b \ell_b^2)^{4/3} [x_1 + (x_2 - x_3)^{1/2}]^{4/3} \geq 0. \quad (7)$$

$$x_4 - t_t^{\min} \geq 0, \quad (8)$$

$$10x_4 - [3.9 \cdot 1.05(.01h_{1t})^{1/2}(.01x_1) + 10K_2] \geq 0, \quad (9)$$

$$10x_4 - [3.9 \cdot 1.05(.01h_{1t})^{1/2}(.01x_2) + 10K_2] \geq 0, \quad (10)$$

$$x_5 - t_m^{\min} \geq 0, \quad (11)$$

$$10x_5 - [3.9 \cdot 1.05(.01h_{1m})^{1/2}(.01x_1) + 10K_2] \geq 0, \quad (12)$$

$$10x_5 - [3.9 \cdot 1.05(.01h_{1m})^{1/2}(.01x_2) + 10K_2] \geq 0, \quad (13)$$

$$x_6 - t_b^{\min} \geq 0, \quad (14)$$

$$10x_6 - [3.9 \cdot 1.05(.01h_{1b})^{1/2}(.01x_1) + 10K_2] \geq 0, \quad (15)$$

$$10x_6 - [3.9 \cdot 1.05(.01h_{1b})^{1/2}(.01x_2) + 10K_2] \geq 0. \quad (16)$$

$$x_i \geq 0; i=1, 2, \text{ and } 3.$$

The listing of the problem parameters and their values is in Table 14.

4.3.2. Computer listing of the symbolic code. (See Figure 5.)

4.3.3. Selected pages from computer output. (See Appendix 1, page 78.)

TABLE 14  
 PROBLEM DESIGN PARAMETERS  
 (SAMPLE PROBLEM 3)

Input Data		
B =	476	cm
$\Gamma$ =	7.85	ton/cm
* $\ell_t$ =	495	cm
* $\ell_m$ =	385	cm
* $\ell_b$ =	315	cm
$h_t$ =	498	cm
$h_m$ =	938	cm
$h_b$ =	1288	cm
$h_{lt}$ =	745	cm
$h_{lm}$ =	1130	cm
$h_{lb}$ =	1445	cm
* $t_t^{\min}$ =	1.05	cm
* $t_m^{\min}$ =	1.05	cm
* $t_b^{\min}$ =	1.05	cm
e =	0.8	cm
$k_1$ =	$6.94 \times 10^{-8}$	cm <sup>-1</sup>
* $k_2$ =	0.15	cm
$h_a$ =	250	cm

\*Parameters marked with an asterisk are selected for sensitivity analysis.



```
//AAAAAA JOB (5660,R,4,4),ABG
// EXEC FORG6,DSN='OR5660.ABG',PROG=MAIN
//SYSLIB DD
//      DD
//      DD DSN=GWU.FORTLIB,DISP=SHR
//GO.FT05F001 DD *
```

```
0.00001      100.      .000001      4.0      800.      19      6      0
      3      2      2      1      1      1      1      2      1      1
EXAMPLE-3 : SOLUTION AND SESITIVITY ANALYSIS OF CORRUGATED BULKHEAD
      6      45      19      0      1      15
```

```
V      X1      45.8
V      X2      43.2
V      X3      30.5
V      X4      1.2
V      X5      1.2
V      X6      1.3
P      LT      495.
P      LM      385.
P      LB      315.
P      TTM      1.05
P      TMM      1.05
P      TBM      1.05
P      K2      0.15
D      E      .8
D      GB      3736.6
D      KHT      0.00003456
D      KHM      0.00006510
D      KHB      0.00008938
D      CT      .011177
D      CM      .013765
D      CB      .015566
S      Z1      X2 POW      2.
S      Z1      X3 POW      2.      -1.
S      Z2      X1 IDN
S      Z2      Z1 POW      .5
Q      Z3      X3 LIN      .5
Q      Z3      X4 IDN
Q      Z4      X3 LIN      .5
Q      Z4      X5 IDN
Q      Z5      X3 LIN      .5
Q      Z5      X6 IDN
S      Z6      X2 LIN      0.33333333
S      Z6      X1 LIN      E
Q      Z7      Z3 IDN
Q      Z7      Z6 IDN
Q      Z8      Z4 IDN
Q      Z8      Z6 IDN
Q      Z9      Z5 IDN
Q      Z9      Z6 IDN
S      Z10     Z2 LIN      KHT
S      Z11     Z2 LIN      KHM
S      Z12     Z2 LIN      KHB
S      Z13     Z10 LIN      LT
S      Z14     Z11 LIN      LM
S      Z15     Z12 LIN      LB
```

Figure 5.--Computer listing of sample problem 3.

S	Z16	Z13	LIN	LT	
S	Z17	Z14	LIN	LM	
S	Z18	Z15	LIN	LB	
S	G1	X2	IDN		
S	G1	X3	LIN	-1.	
S	G2	Z7	IDN		
S	G2	Z16	LIN	-1.	
S	G3	Z8	IDN		
S	G3	Z17	LIN	-1.	
S	G4	Z9	IDN		
S	G4	Z18	LIN	-1.	
Q	G5	Z7	IDN		
Q	G5	X3	LIN	.5	
S	G5	Z16	POW	1.33333333	-2.2
Q	G6	Z8	IDN		
Q	G6	X3	LIN	.5	
S	G6	Z17	POW	1.33333333	-2.2
Q	G7	Z9	IDN		
Q	G7	X3	LIN	.5	
S	G7	Z18	POW	1.33333333	-2.2
S	G8	X4	LIN	-1.	TTM
S	G9	X4	LIN	-1.	
S	G9	X1	LIN	CT	K2
S	G10	X4	LIN	-1.	
S	G10	X2	LIN	CT	K2
S	G11	X5	LIN	-1.	TMM
S	G12	X5	LIN	-1.	
S	G12	X1	LIN	CM	K2
S	G13	X5	LIN	-1.	
S	G13	X2	LIN	CM	K2
S	G14	X6	LIN	-1.	TBM
S	G15	X6	LIN	-1.	
S	G15	X1	LIN	CB	K2
S	G16	X6	LIN	-1.	
S	G16	X2	LIN	CB	K2
S	Z19	X1	IDN		
S	Z19	X2	IDN		
S	Z20	Z19	LIN	GB	
S	Z21	X4	LIN	LT	
S	Z21	X5	LIN	LM	
S	Z21	X6	LIN	LB	
Q	Z22	Z20	IDN		
Q	Z22	Z21	IDN		
Q	OBJ	Z22	IDN		
Q	OBJ	Z2	POW	-1.	
L	X1	X3		0.	
L	G1	G7		0.	
U	G8	G16		0.	
S					

.001 .001

//

Figure 5.--continued

## 5. General Description of the New Subroutines

The following is a list of the names of the subroutines that comprise the SYMBOLIC FACTORABLE SUMT model in alphabetical order:

BODY	OUTPUT
CHCKER	PARSEN*
CONVRG	PEVALU
DLPEN*	PUNCH
ESTIM	REJECT
EVALU	RESTNT**
EXDELP*	RHOCOM
FEAS	SECORD
FINAL	SENS*
GRAD	STORE
GRAD1**	SYMPUT*
GVALU*	TCHECK
INVERS	TIMEC
MAIN	VALU*
MATRIX**	XMOVE
OPT	

Subroutines marked with a single asterisk (\*) are the new subroutines that have been added to the SUMT-Version 4 model in its various developmental stages to result in SYMBOLIC SENSITIVITY SUMT model. Subroutines marked with a double asterisk (\*\*) correspond to the user's subroutine in SUMT-Version 4.

In order to provide problem function values, their gradients and Hessian matrices, users had to code the latter subroutines in SUMT-Version 4, while here, in conjunction with the new subroutines, they are modified to serve the same purpose internally. The remainder of the subroutines virtually serve the same purposes as in SUMT-Version 4.

For brevity, a general description of the new subroutines is provided here. Readers interested in general descriptions of the remaining subroutines must refer to [1].

The new subroutines fall into two categories,

- (i) those developed to relax the user from coding old users' subroutines--these subroutines are SYMPUT, VALU, and GVALU; and

- (ii) those developed to perform sensitivity and elasticity calculations--these subroutines are PARSEN, SENS, EXDELP, and DLPEN.

Following is a brief and general description of the role of each of these subroutines.

#### *SYMPUT*

Reads in and stores in arrays the coded problem in symbolic factorable format such that problem functions can be synthesized with the aid of these arrays. The title, SYMPUT parameter, and flag cards are also read in by this subroutine. It prints out a modified listing of the coded problem if desired. As described in Section 3 (Tables 7 and 8), this subroutine selects F or A format internally, while reading the coded problem, for numeric and alphanumeric inputs, respectively. This is made possible with the aid of subroutine REREAD, which is a special purpose George Washington University FORTRAN Library routine. Following is a brief description of this subroutine.

"Subroutine Name: REREAD - permits reformatting of a record.

Description: REREAD patches the FORTRAN I/O routine so that upon execution it examines every read for a data set reference number (device code) of 99. If 99 is not found normal reading takes place. If 99 is found reading does not take place and the record read by the last normal read is formatted according to the new format. The call needs to be issued only once for the program.

Usage: CALL REREAD

Example: DIMENSION A(10),B(20)  
CALL REREAD  
.....  
.....  
READ (5,100) A This will read 10 variables  
on a card.  
READ (99,101) B This will reformat the card  
into 20A1.  
.....  
.....  
100 FORMAT(10F2.0)  
101 FORMAT(20A1)."



*VALU*

Evaluates the functional values, first and second derivatives of single variable functions as required by GVALU. These results are transferred to subroutine GVALU via common blocks to reconstruct the problem function values, gradients, and Hessian matrices.

*GVALU*

Using the information stored in subroutine SYMPUT regarding the coded problem, via common blocks, GVALU reconstructs the problem function values, gradients, and Hessian matrices with the aid of subroutine VALU.

*PARSEN*

Acts as a coordinating routine to conduct sensitivity analysis in the solution vector of each subproblem. It also calculates the first order sensitivity information, elasticities of solution vector with respect to the problem parameters, and prints them out.

*SENS*

Evaluates the functional values, first and second derivatives of single variable functions, partial derivatives of these functions with respect to function parameters (C1 and C2), and joint partials with respect to parameters and function variables as required by subroutine EXDELP. This information is transferred to subroutine EXDELP for reconstruction of functional values, gradients, Hessian matrices of the problem functions, and related partial derivatives and cross partial matrices required for sensitivity analysis.

*EXDELP*

Using the information stored in the subroutine SYMPUT about the coded problem, and utilizing the information calculated in SENS, via common blocks, reconstructs the functional values, gradients, Hessian matrices of the problem functions, their partials with respect to the problem parameters, and joint partials with respect to the problem variables and parameters.

*DLPEN*

Using the information calculated in EXDELP via common blocks, uses the sensitivity formulas given in Appendix 2, to calculate the zeroth order sensitivity of the solution vector with respect to the problem parameters.

An important point worth noting here is that the subroutines SYM-PUT, VALU, GVALU, SENS, and EXDELP are quite general subroutines that can be interfaced with any nonlinear programming algorithm which uses the first or the first and second derivative methods.

Subroutines comprising SYMBOLIC FACTORABLE SUMT are currently dimensioned to handle problems of the following maximum scale:

- total number of CVFs  $\leq 100$
- number of original variables  $\leq 30$
- total number of parameters  $\leq 75$
- number of sensitivity parameters  $\leq 15$
- total number of original problem constraints  $\leq 200$  .

However, if computer capacity permits, they can readily be redimensioned to solve problems of higher dimensions. All of these subroutines are separately filed in the TED facility [9] at The George Washington University under their corresponding names. They may be accessed from the above source using account number 55599 and the proper password.

## REFERENCES

- [1] MYLANDER, W. C., R. L. HOLMES, and G. P. McCORMICK (1971). A guide to SUMT-Version 4: The computer program implementing the sequential unconstrained minimization technique for nonlinear programming. RAC Paper RAC-P-63, Research Analysis Corporation, McLean, Virginia.
- [2] McCORMICK, G. P. (1974). A mini-manual for use of the SUMT computer program and the factorable programming language. Technical Report SOL 74-15, System Optimization Laboratory, Department of Operations Research, Stanford University, Stanford, California.
- [3] deSILVA, A. H. and G. P. McCORMICK (1977). An interim manual for using the SUMT algorithm in conjunction with the symbolic factorable programming language. Draft paper.
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- [8] FIACCO, A. V. and A. GHAEMI (1979). Sensitivity analysis of a non-linear structural design problem. In preparation.
- [9] TED FACILITY USER'S GUIDE, V2M0 (1977). Center for Academic and Administrative Computing, George Washington University.
- [10] FIACCO, A. V. and G. P. McCORMICK (1968). *Nonlinear Programming: Sequential Unconstrained Minimization Techniques*. John Wiley and Sons, Inc., New York.



APPENDIX 1

Annotated Computer Output for Solution  
of Sample Problem 1 (pages 50-68) and  
Selected Pages of Computer Solution  
Output for Sample Problems 2  
(pages 69-77) and 3  
(pages 78-84)

## EXAMPLE-1 : SOLUTION AND SENSITIVITY ANALYSIS

SYMPUT PARAMETER CARD

N= 2 NN= 6 M= 4 NZ= 0 ITEST= 1 IPR= 0 ITES= 0 NSUIT= 0 IID= 6

TYPE NAME	TYPE #	TYPE DESCRIP.	VAR. NAME	NUMBR. ASSND.	ARG. VARIAB. NAME	NUMBR.	FUNCT. TYPE	NUMBR. ASSND.	CONSTANT C1	CONSTANT C2
V	ORIG. VARIABLE	X1	1	0.200000	01					
V	ORIG. VARIABLE	X2	2	0.200000	01					
P	SENSITIVITY PAR	C5				1		0.100000	01	
P	SENSITIVITY PAR	C6				2		-0.150000	01	
D	DATA DEFINITION	C1						0.0		
D	DATA DEFINITION	C2						0.0		
D	DATA DEFINITION	C3						0.300000	01	
D	DATA DEFINITION	C4						0.150000	01	
S	SEPARABL	X3	3			1	SIN	20	0.500000	00
S	SEPARABL	X3	3			2	LIN	3	0.100000	01
S	SEPARABL	X4	4			2	LIN	3	0.100000	01
S	SEPARABL	X4	4			1	LNG	15	0.100000	01
Q	QUADRATC	X5	5			1	LIN	3	0.100000	01
Q	QUADRATC	X5	5			2	LIN	3	0.100000	01
S	SEPARABL	X6	6			1	QDR	33	0.100000	01
S	SEPARABL	X6	6			2	QDR	33	0.100000	01

ARRAY OF CONSTANTS CCL(.,.)

\*\*\*\*\*

LOCATION VALUE

1	0.100000	01
2	-0.150000	01
3	0.0	
4	0.0	
5	0.300000	01
6	0.150000	01
7	0.0	
8	0.100000	01
9	0.500000	00
10	-0.500000	01
11	0.100000	02

BOUNDS: NCONC(.,.) : LOWER BOUND HAS MINUS SIGN

CVF CVF# VALUE OF BOUND LOCATION IN CCL(.,.)

1	X3	3	0.0	3
2	X4	4	0.0	-4
3	X5	5	0.300000	01
4	X1	1	0.150000	01

RELATIVE LOCATION: ITAB(.,.,.)

CVF# SEPARABLE QUADRATIC

3	1	1
4	3	1
5	5	1
6	5	2
7	7	2

- 51 -

ARG	ARG#	TRANSF	C1	LOC.IN	C2	LOC.IN	CCI
1	X1	1	20	SIN	0.5000	00	9
2	X2	2	3	LIN	0.1000	01	1
3	X2	2	3	LIN	0.1000	01	8
4	X1	1	15	LNG	0.1000	01	8
5	X1	1	33	QDR	0.1000	01	8
6	X2	2	33	QDR	0.1000	01	8

QUADRATIC DEFINITIONS : IUADTB(.,.,.) : EACH LINE REFERS TO A PAIR OF ARGUMENTS

\*\*\*\*\* FIRST ARGUMENT \*\*\*\*\* SECOND ARGUMENT \*\*\*\*\*

ARG	ARG#	TRANSF	C1	LOC.IN	C2	LOC.IN	CCI
1	X1	1	3	LIN	0.1000	01	8
2	X2	2	3	LIN	0.1000	01	8
3	X2	2	3	LIN	0.1000	01	8
4	X1	1	15	LNG	0.1000	01	8
5	X1	1	33	QDR	0.1000	01	8
6	X2	2	33	QDR	0.1000	01	8

7

8

9 N= 2 M= 4 MZ= 0  
 MAX. TIME= 0.5000000 03 R= 0.1000000 03 RATIO= 0.1000000 02 EPSILON= 0.1000000-04 THETA= 0.1000000-04  
 10 OPTIONS SELECTED 1 1 1 1 2 1 1  
 11 TOLERANCES 0.1000000-02 0.1000000-02  
 12 SECOND SET OF OPTIONS  
 13 F= 0.6500000 02 P= 0.0 G= 0.0  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;  
 X1= 0.20000000 01 X2= 0.20000000 01  
 THE CONSTRAINT VALUES  
 0.2207350 01 0.9602792 00 0.1000000 01 0.5000000 00  
 14 \*\*\*\*\*THE FEASIBLE STARTING POINT TO BE USED IS ...  
 F= 0.6500000 02 P= 0.0 G= 0.0  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;  
 X1= 0.20000000 01 X2= 0.20000000 01  
 THE CONSTRAINT VALUES  
 0.2207350 01 0.9602792 00 0.1000000 01 0.5000000 00  
 \*\*\*\*\*  
 POINT= 3 DOT1= 0.95411290-05 RHO= 0.1000000 03 MAGNITUDE= 0.31566160-01 PHASE= 2  
 F= 0.46097530 02 P= -0.33441220 03 G= -0.35390250 03  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;  
 X1= 0.37505537D 01 X2= 0.36536666D 01  
 THE CONSTRAINT VALUES  
 0.11163480 01 0.1670811D 01 0.10703270 02 0.22505540 01  
 15 LAGRANGE MULTIPLIERS  
 F= 0.46097530 02 P= -0.33441220 03 G= -0.35390250 03 RHO= -0.38050970 03 M= 0.0  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;  
 X1= 0.37505537D 01 X2= 0.36536666D 01  
 THE CONSTRAINT VALUES  
 0.89577840 02 0.5985116D 02 0.93429370 01 0.44433510 02

SENSITIVITY OF SOLUTION VECTOR WRT PROBLEM PARAMETERS  
 \*\*\*\*\*

PROBLEM PARAMETERS ARE  
 \*\*\*\*\*

C5 C6

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.465140 00 0.314990 00

RESPECTIVE ELASTICITIES ARE :



-0.12432D 00 -0.12598D 00

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.27637D 01 -0.48034D 00

RESPECTIVE ELASTICITIES ARE :

-0.75642D 00 0.19720D 00

APPARENTLY ROUND OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.

\*\*\*\*\*  
 POINT= 6 DOTT= 0.1471231D-04 RHO= 0.1000000D 02 MAGNITUDE= 0.2037057D-01 PHASE= 2  
 F= 0.3680283D 02 P= 0.3288146D 01 G= -0.3197171D 01 RSIGMA= -0.3351468D 02 H= 0.0  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.43303384D 01 X2= 0.31581859D 01  
 THE CONSTRAINT VALUES  
 0.9843133D 00 0.9597174D 00 0.1067601D 02 0.2830338D 01

1ST ORDER ESTIMATES

F= 0.3584189D 02 P= 0.4081041D 02 G= 0.3577009D 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.43947589D 01 X2= 0.31031325D 01  
 THE CONSTRAINT VALUES  
 0.9470465D 00 0.8825135D 00 0.1063752D 02 0.2894759D 01

LAGRANGE MULTIPLIERS

F= 0.3584189D 02 P= 0.4081041D 02 G= 0.3577009D 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.43947589D 01 X2= 0.31031325D 01  
 THE CONSTRAINT VALUES  
 0.1015937D 02 0.1041973D 02 0.9366792D 00 0.3533146D 01

SENSITIVITY OF SOLUTION VECTOR WRT PROBLEM PARAMETERS

PROBLEM PARAMETERS ARE

C5 C6

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.87601D 00 0.48925D 00

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.92166D 00 0.50861D 00

RESPECTIVE ELASTICITIES ARE :

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.13339D 01 -0.85640D 00

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.11750D 01 -0.89818D 00

RESPECTIVE ELASTICITIES ARE :

-0.37206D 00 0.42660D 00

\*\*\*\*\*  
 POINT= 10 DOTT= 0.58869590-05 RHO= 0.1000000D 01 MAGNITUDE= 0.1815338D-01 PHASE= 2  
 F= 0.2673064D 02 P= 0.2645202D 02 G= 0.2273064D 02 RSIGMA= -0.2786173D 00 H= 0.0  
 THE CURRENT VALUE OF THE PROB-MIL.E.X(11) VARS.ARE;

X1= 0.50790812D 01 X2= 0.25859375D 01  
 THE CONSTRAINT VALUES  
 0.2457370D 00 0.1482419D 00 0.1013419D 02 0.3579081D 01

2ND ORDER ESTIMATES  
 F= 0.2562837D 02 P= 0.2890675D 02 G= 0.2561151D 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-MIL.E.X(11) VARS.ARE;

X1= 0.51700276D 01 X2= 0.25164879D 01  
 THE CONSTRAINT VALUES  
 0.1249347D 00 0.5217083D-01 0.1001031D 02 0.3670028D 01

1ST ORDER ESTIMATES  
 F= 0.2572115D 02 P= 0.2902579D 02 G= 0.2561151D 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-MIL.E.X(11) VARS.ARE;

X1= 0.51622749D 01 X2= 0.25223543D 01  
 THE CONSTRAINT VALUES  
 0.1355048D 00 0.6028828D-01 0.1002109D 02 0.3662275D 01

LAGRANGE MULTIPLIERS  
 F= 0.2572115D 02 P= 0.2902579D 02 G= 0.2561151D 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-MIL.E.X(11) VARS.ARE;

X1= 0.51622749D 01 X2= 0.25223543D 01  
 THE CONSTRAINT VALUES  
 0.4069391D 01 0.6745731D 01 0.9867590D-01 0.2794013D 00

SENSITIVITY OF SOLUTION VECTOR WRT PROBLEM PARAMETERS  
 \*\*\*\*\*

PROBLEM PARAMETERS ARE  
 \*\*\*\*\*

C5 C6

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.101120 01 0.654100 00

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.102620 01 0.672420 00

RESPECTIVE ELASTICITIES ARE :

-0.202040 00 -0.198580 00

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.403270 00 -0.136720 01

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.299060 00 -0.142390 01

RESPECTIVE ELASTICITIES ARE :

-0.115960 00 0.825950 00

APPARENTLY ROUND OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.

\*\*\*\*\*

POINT= 13 DOT1= 0.65923700-04

F= 0.25036430 02 P= 0.25449520 02

THE CURRENT VALUE OF THE PROB-MII.E.X(II) VARS.ARE:

X1= 0.522503540 01 X2= 0.249556520 01

THE CONSTRAINT VALUES

0.27980360-01 0.15355600-01 0.10039570 02 0.37250950 01

2ND ORDER ESTIMATES

F= 0.24843120 02

P= 0.25300880 02

THE CURRENT VALUE OF THE PROB-MII.E.X(II) VARS.ARE:

X1= 0.524211760 01 X2= 0.248515180 01

THE CONSTRAINT VALUES

0.15650420-02 0.63482790-04 0.10027460 02 0.37421180 01

1ST ORDER ESTIMATES

F= 0.24851820 02

P= 0.25338130 02

THE CURRENT VALUE OF THE PROB-MII.E.X(II) VARS.ARE:

X1= 0.524131920 01 X2= 0.248552380 01

THE CONSTRAINT VALUES

0.29245140-02 0.66399040-03 0.10027420 02 0.37413190 01

LAGRANGE MULTIPLIERS

F= 0.24851820 02

P= 0.25338130 02

THE CURRENT VALUE OF THE PROB-MII.E.X(II) VARS.ARE:

X1= 0.524131920 01 X2= 0.248552380 01

THE CONSTRAINT VALUES

0.35739360 01 0.65122810 01 0.99605900-02 0.26844950-01

MAGNITUDE= 0.20271320 00  
RSIGMA= 0.41309010 00  
H= 0.0  
PHASE= 2



SENSITIVITY OF SOLUTION VECTOR WRT PROBLEM PARAMETERS  
\*\*\*\*\*

PROBLEM PARAMETERS ARE  
\*\*\*\*\*

C5 C6

THE ESTIMATES OF ZEROTH ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
-----

-0.96309D 00 0.65477D 00

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
-----

-0.95775D 00 0.65484D 00

RESPECTIVE ELASTICITIES ARE :  
\*\*\*\*\*

-0.18330D 00 -0.18799D 00

THE ESTIMATES OF ZEROTH ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
-----

-0.28617D 00 -0.14148D 01

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
-----

-0.27316D 00 -0.14201D 01

RESPECTIVE ELASTICITIES ARE :  
\*\*\*\*\*

-0.10946D 00 0.85359D 00

\*\*\*\*\*  
POINT= 15 DOTT= 0.5655324D-06 RHO= 0.1000000D-01 MAGNITUDE= 0.6452718D-01 PHASE= 2  
F= 0.2485718D 02 P= 0.2494437D 02 G= 0.2481718D 02 RSIGMA= 0.8718882D-01 H= 0.0  
THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.5241000D 01 X2= 0.24863049D 01  
THE CONSTRAINT VALUES  
0.2835304D-02 0.1536448D-02 0.1003072D 02 0.3741000D 01

2ND ORDER ESTIMATES

F= 0.2483716D 02 P= 0.2488370D 02 G= 0.2483727D 02 RSIGMA= 0.0 H= 0.0  
THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.52427819D 01 X2= 0.24852735D 01  
THE CONSTRAINT VALUES  
0.2425232D-05 -0.4847195D-05 0.1002975D 02 0.3742782D 01

1ST ORDER ESTIMATES

F= 0.2483731D 02 P= 0.2488824D 02 G= 0.2483727D 02 RSIGMA= 0.0 H= 0.0  
THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.52427672D 01 X2= 0.24852760D 01  
THE CONSTRAINT VALUES



- 57 -

LAGRANGE MULTIPLIERS  
 F= 0.248371D 02 P= 0.2488824D 02 G= 0.2483727D 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE:

X1= 0.52427672D 01 X2= 0.24852760D 01  
 THE CONSTRAINT VALUES  
 0.3526958D 01 0.6508518D 01 0.9969370D 03 0.2673082D 02

SENSITIVITY OF SOLUTION VECTOR WRT PROBLEM PARAMETERS  
 \*\*\*\*\*

PROBLEM PARAMETERS ARE  
 \*\*\*\*\*

C5 C6

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
 -----

-0.99476D 00 0.68074D 00

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
 -----

-0.99828D 00 0.68363D 00

RESPECTIVE ELASTICITIES ARE :  
 \*\*\*\*\*

-0.19047D 00 -0.19566D 00

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
 -----

-0.28570D 00 -0.14761D 01

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
 -----

-0.28565D 00 -0.14830D 01

RESPECTIVE ELASTICITIES ARE :  
 \*\*\*\*\*

-0.11489D 00 0.89468D 00

\*\*\*\*\*  
 POINT= 17 DOT= 0.438981D 06 RHO= 0.100000D 02 MAGNITUDE= 0.1529174D 00 PHASE= 2  
 F= 0.2483918D 02 P= 0.2485251D 02 G= 0.2483518D 02 RSIGMA= 0.1332714D 01 H= 0.0  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE:

X1= 0.52426029D 01 X2= 0.24853800D 01  
 THE CONSTRAINT VALUES  
 0.2860025D 03 0.1528636D 03 0.1002986D 02 0.3742603D 01

2ND ORDER ESTIMATES  
 F= 0.2483718D 02 P= 0.2484184D 02 G= 0.2483716D 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE:

- 58 -

THE CONSTRAINT VALUES  
0.11047300-06 -0.88662050-06 0.10029770 02 0.37427810 01

1ST ORDER ESTIMATES  
F= 0.24837180 02 P= 0.24842300 02 G= 0.24837180 02 RSIGMA= 0.0 H= 0.0  
THE CURRENT VALUE OF THE PROB-MT.E.X(1) VARS.ARE;

X1= 0.52427810 01 X2= 0.248527730 01  
THE CONSTRAINT VALUES  
0.42589590-06 -0.85934840-06 0.10029760 02 0.37427810 01

LAGRANGE MULTIPLIERS  
F= 0.24837180 02 P= 0.24842300 02 G= 0.24837180 02 RSIGMA= 0.0 H= 0.0  
THE CURRENT VALUE OF THE PROB-MT.E.X(1) VARS.ARE;

X1= 0.52427810 01 X2= 0.248527730 01  
THE CONSTRAINT VALUES  
0.35210960 01 0.65417790 01 0.99702280-04 0.26719370-03

SENSITIVITY OF SOLUTION VECTOR WRT PROBLEM PARAMETERS  
\*\*\*\*\*

PROBLEM PARAMETERS ARE  
\*\*\*\*\*

C5 C6

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
-----

-0.991850 00 0.710220 00

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
-----

-0.991530 00 0.713500 00

RESPECTIVE ELASTICITIES ARE :  
\*\*\*\*\*

-0.189130 00 -0.204140 00

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
-----

-0.283890 00 -0.154060 01

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
-----

-0.283690 00 -0.154780 01

RESPECTIVE ELASTICITIES ARE :  
\*\*\*\*\*

-0.114140 00 0.934130 00

\*\*\*\*\*  
POINT= 19 DOT= 0.80839870-06 RMU= 0.10000000-03 MAGNITUDE= 0.82014770 00 PHASE= 2  
\*\*\*\*\*  
C= 0.2624000 02 SIGMA= 0.17000000-02 H= 0.0

- 59 -

THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.52427628D 01 X2= 0.24852884D 01  
 THE CONSTRAINT VALUES  
 0.2882452D-04 0.1542979D-04 0.1002978D 02 0.3742763D 01

2ND ORDER ESTIMATES  
 F= 0.2483719D 02 P= 0.2483765D 02 G= 0.2483719D 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.52427806D 01 X2= 0.24852782D 01  
 THE CONSTRAINT VALUES  
 0.4708845D-06 0.1697422D-06 0.1002977D 02 0.3742781D 01

1ST ORDER ESTIMATES  
 F= 0.2483719D 02 P= 0.2483770D 02 G= 0.2483719D 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.52427806D 01 X2= 0.24852782D 01  
 THE CONSTRAINT VALUES  
 0.4704346D-06 0.1594513D-06 0.1002977D 02 0.3742781D 01

LAGRANGE MULTIPLIERS  
 F= 0.2483719D 02 P= 0.2483770D 02 G= 0.2483719D 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.52427806D 01 X2= 0.24852782D 01  
 THE CONSTRAINT VALUES  
 0.3469268D 01 0.6480970D 01 0.9970311D-05 0.2671823D-04

SENSITIVITY OF SOLUTION VECTOR WRT PROBLEM PARAMETERS  
 \*\*\*\*\*

PROBLEM PARAMETERS ARE  
 \*\*\*\*\*

C5

C6

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
 -----

-0.81119D 00 0.65820D 00

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
 -----

-0.79112D 00 0.65242D 00

RESPECTIVE ELASTICITIES ARE :  
 =====

-0.15090D 00 -0.18666D 00

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
 -----

-0.23210D 00 -0.14278D 01

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
 -----



- 60 -

-0.226340 00 -0.141530 01

RESPECTIVE ELASTICITIES ARE :  
=====

-0.910740-01 0.854200 00

\*\*\*\*\*

POINT= 21 DOTI= 0.63905140-06 RHU= 0.10000000-04 MAGNITUDE= 0.25509340 01 PHASE= 2  
 F= 0.24837180 02 P= 0.24837430 02 G= 0.24837170 02 RSIGMA= 0.22514500-03 H= 0.0  
 THE CURRENT VALUE OF THE PROB-MII.E.X(1) ) VARS.ARE:

X1= 0.52427790 01 X2= 0.248527910 01  
 THE CONSTRAINT VALUES  
 0.28367280-05 0.15659290-05 0.10029770 02 0.37427790 01

2ND ORDER ESTIMATES

F= 0.24837180 02 P= 0.24837240 02 G= 0.24837180 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-MII.E.X(1) ) VARS.ARE:

X1= 0.52427800 01 X2= 0.248527810 01  
 THE CONSTRAINT VALUES  
 -0.56079490-07 0.24148130-07 0.10029770 02 0.37427810 01

1ST ORDER ESTIMATES

F= 0.24837180 02 P= 0.24837240 02 G= 0.24837180 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-MII.E.X(1) ) VARS.ARE:

X1= 0.52427800 01 X2= 0.248527810 01  
 THE CONSTRAINT VALUES  
 -0.50814350-07 0.25501160-07 0.10029770 02 0.37427810 01

LAGRANGE MULTIPLIERS

F= 0.24837180 02 P= 0.24837240 02 G= 0.24837180 02 RSIGMA= 0.0 H= 0.0  
 THE CURRENT VALUE OF THE PROB-MII.E.X(1) ) VARS.ARE:

X1= 0.52427800 01 X2= 0.248527810 01  
 THE CONSTRAINT VALUES  
 0.35251880 01 0.63859850 01 0.99703190-06 0.26718110-05

SENSITIVITY OF SOLUTION VECTOR WRT PROBLEM PARAMETERS  
 \*\*\*\*\*

PROBLEM PARAMETERS ARE  
 =====

C5

C6

THE ESTIMATES OF ZEROTH ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
 -----

-0.594650 00 0.529210 00

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
 -----

-0.570590 00 0.514870 00

RESPECTIVE ELASTICITIES ARE :  
 -----



- 61 -

-0.108830 00 -0.147310 00

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
-----

-0.170130 00 -0.114800 01

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
-----

-0.163250 00 -0.111690 01

RESPECTIVE ELASTICITIES ARE :  
=====

-0.656870-01 0.674120 00

SYNPUT PARAMETER CARD  
 N= 2 NN= 6 M= 4 MZ= 0 ITEST= 0 IPR= 0 ITES= 0 NSUIT= 0 IIO= 6

## EXAMPLE-1 : SOLUTION AND SENSITIVITY ANALYSTS

TYPE NAME	TYPE #	DESCRIP.	VAR. NAME	NUMBR. ASSND.	APG. VAR. NAME	NUMBR.	FUNCT. TYPE	NUMBR. ASSND.	CONSTANT C1	CONSTANT C2
V	1	ORIG. VARIABLE	X1	1	0.200000 01					
V	2	ORIG. VARIABLE	X2	2	0.200000 01					
P	5	SENSITIVITY PAR	C5			1		0.100000 01		
P	6	SENSITIVITY PAR	C6			2		-0.150000 01		
D	1	DATA DEFINITION	C1					0.0		
D	2	DATA DEFINITION	C2					0.0		
D	3	DATA DEFINITION	C3					0.3000000000 01		
D	4	DATA DEFINITION	C4					0.1500000000 01		
S	3	SEPARABL	X3	3		1	SIN	20 0.5000000000 00	-0.5000000000 01	
S	3	SEPARABL	X3	3		2	LIN	3 0.1000000000 01	0.0	
S	3	SEPARABL	X4	4		2	LIN	3 0.1000000000 01	0.0	
S	3	SEPARABL	X4	4		1	LNG	15 0.1000000000 01	-0.1500000000 01	
Q	4	QUADRATIC	X5	5		1	LIN	3 0.1000000000 01	0.0	
Q	4	QUADRATIC	X5	5		2	LIN	3 0.1000000000 01	0.0	
S	3	SEPARABL	X6	6		1	QDR	33 0.1000000000 01	0.1000000000 02	
S	3	SEPARABL	X6	6		2	QDR	33 0.1000000000 01	0.1000000000 01	

ARRAY OF CONSTANTS CCL(.,.)  
 \*\*\*\*\*

LOCATION	VALUE
1	0.100000 01
2	-0.150000 01
3	0.0
4	0.0
5	0.300000 01
6	0.150000 01
7	0.0
8	0.100000 01
9	0.500000 00
10	-0.500000 01
11	0.100000 02

BOUNDS:NCUNC(.,.) : LOWER BOUND HAS MINUS SIGN

CVF	CVF#	VALUE OF BOUND	LOCATION IN CCL(.,.)
1	X3	3 0.0	3
2	X4	4 0.0	-4
3	X5	5 0.30000000 01	-5
4	X1	1 0.15000000 01	-6

RELATIVE LOCATION:ITAB(.,.)  
 CVF# SEPARABLE QUADRATIC

3	1	1
4	3	1
5	5	1
6	5	2
7	7	2

## SEPARABLE DEFINITIONS : IEP(TAB1,...)

ARG	ARG#	TRANSF	C1	LUC.IN	C2	LOC.IN	CCI
1	X1	1	20	SIN	0.5000	00	9
2	X2	2	3	LIN	0.1000	01	10
3	X2	2	3	LIN	0.1000	01	1
4	X1	1	15	LNG	0.1000	01	8
5	X1	1	33	QDR	0.1000	01	8
6	X2	2	33	QDR	0.1000	01	8

## QUADRATIC DEFINITIONS : IUA(TB1,...) : EACH LINE REFERS TO A PAIR OF ARGUMENTS

ARG	ARG#	TRANSF	C1	LUC.IN	C2	LOC.IN	CCI	ARG	ARG#	TRANSF	C1	LUC.IN	C2	LOC.IN	CCI				
1	X1	1	3	LIN	0.1000	01	8	0.0	7	*	X2	2	3	LIN	0.1000	01	8	0.0	7

- 64 -

## NONLINEAR PROGRAMMING ROUTINE-SUMT VERSION 4 08/10/71

N= 2 M= 4 MZ= 0

MAX. TIME= 0.5000000 03 R= 0.1000000 03 RATIO= 0.1000000 02 EPSILON= 0.10000000-04 THETA= 0.10000000-04

## OPTIONS SELECTED

1 1 1 1 2 1 1

## TOLERANCES

0.10000000-02 0.10000000-02

## SECOND SET OF OPTIONS

1 1

F= 0.6500000 02 P= 0.0 G= 0.0  
THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.20000000 01 X2= 0.20000000 01

## THE CONSTRAINT VALUES

0.22073550 01 0.96027920 00 0.10000000 01 0.50000000 00

\*\*\*\*\*THE FEASIBLE STARTING POINT TO BE USED IS ...

F= 0.65000000 02 P= 0.0 G= 0.0  
THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

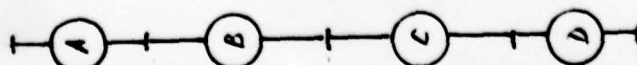
X1= 0.20000000 01 X2= 0.20000000 01

## THE CONSTRAINT VALUES

0.22073550 01 0.96027920 00 0.10000000 01 0.50000000 00

1 0.20000000 01  
2 0.20000000 01  
3 0.20000000 01  
4 -0.220735490 01  
5 0.960279230 00  
6 0.400000000 01  
0.650000000 021 -0.1350760 01 0.1000000 01  
2 -0.7500000 00 0.1000000 01  
3 0.2000000 01 0.2000000 01  
4 -0.1600000 02 0.2000000 011 0.1051840 01 0.0  
2 0.3750000 00 0.0  
3 0.0 0.0  
4 0.2000000 01 0.2000000 01

0.1000000 01





## EXAMPLE-1 : SOLUTION AND SENSITIVITY ANALYSIS

SYMPUT PARAMETER CARD  
N= 2 NN= 5 M= 4 MZ= 0 ITEST= 1 IPR= 0 ITES= 2 NSMIT= 0 IID= 6

TYPE NAME	TYPE #	TYPE DESCRIP.	VAR. NAME	NUMBR. ASSND.	ARG. VARIAB. NAME	NUMBR.	FUNCT. TYPE	NUMBR. ASSND.	CONSTANT C1	CONSTANT C2
V	1	ORIG. VARIABLE	X1	1	0.200000 01					
V	2	ORIG. VARIABLE	X2	2	0.200000 01					
P	1	SENSITIVITY PAR	C5					1	0.100000 01	
P	2	SENSITIVITY PAR	C6					2	-0.150000 01	
D	1	DATA DEFINITION	C1						0.0	
D	2	DATA DEFINITION	C2						0.0	
D	3	DATA DEFINITION	C3						0.300000000 01	
D	4	DATA DEFINITION	C4						0.150000000 01	
S	3	SEPARABL	X3	3		X1	1	20	0.50000000 00	-500000000 01
S	3	SEPARABL	X3	3		X2	2	3	0.10000000 01	0.0
S	3	SEPARABL	X4	4		X2	2		C5	
S	3	SEPARABL	X4	4		X1	1	15	0.10000000 01	-150000000 01
Q	4	QUADRATIC	X5	5		X1	1	3	0.10000000 01	0.0
Q	4	QUADRATIC	X5	5		X2	2	3	0.10000000 01	0.0
S	3	SEPARABL	X6	6		X1	1	33	0.10000000 01	0.100000000 02
S	3	SEPARABL	X6	6		X2	2	33	0.10000000 01	0.100000000 01

ARRAY OF CONSTANTS CC1(..)  
\*\*\*\*\*

LOCATION	VALUE
1	0.100000 01
2	-0.150000 01
3	0.0
4	0.0
5	0.300000 01
6	0.150000 01
7	0.0
8	0.100000 01
9	0.500000 00
10	-0.500000 01
11	0.100000 02

BOUND:NC(....) : LOWER BOUND HAS MINUS SIGN

CVF	CVF#	VALUE OF BOUND	LOCATION IN CC1(..)
1	X3	3	0.0
2	X4	4	0.0
3	X5	5	0.300000 01
4	X1	1	0.150000 01

RELATIVE LOCATION:ITAB(....)  
CVF# SEPARABLE QUADRATIC

3	1	1
4	3	1
5	5	1
6	5	2
7	7	2

- 66 -

```

SEPARABLE DEFINITIONS : IEPFAB(.,.,.)
ARG  ARG# TRANSF C1 LOC.IN C2 LOC.IN
1  X1  1  20  SIN  0.5000 0J  5  -.5000 01 10
2  X2  2  3  LIN  0.1000 01 1  0.0  7
3  X2  2  3  LIN  0.1000 01 8  0.0  7
4  X1  1  15  LNC  0.1000 01 8  -.1500 01 2
5  X1  1  33  QDR  0.1000 01 8  0.1000 02 11
6  X2  2  33  QDR  0.1000 01 8  0.1000 01 8

```

QUADRATIC DEFINITIONS : IJADTB(.,.,.) : EACH LINE REFERS TO A PAIR OF ARGUMENTS

```

***** FIRST ARGUMENT *****
ARG  ARG# TRANSF C1 LOC.IN C2 LOC.IN CCI
1  X1  1  3  LIN  0.1000 01 8  0.0  7 *
2  X2  2  3  LIN  0.1000 01 8  0.0  7
3  X2  2  3  LIN  0.1000 01 8  0.0  7
***** SECOND ARGUMENT *****
ARG  ARG# TRANSF C1 LOC.IN C2 LOC.IN
1  X1  1  3  LIN  0.1000 01 8  0.0  7

```

## NONLINEAR PROGRAMMING ROUTINE-SUMT VERSION 4 08/10/71

N= 2 M= 4 M2= 0  
 MAX. TIME= 0.5000000 03 R= 0.1000000 03 RATIO= 0.1000000 02 EPSILON= 0.1000000D-04 THETA= 0.1000000D-04

OPTIONS SELECTED  
 3 2 1 1 1 1 1 2 1 1

TOLERANCES  
 0.1000000D-02 0.1000000D-02

## SECOND SET OF OPTIONS

1  
 F= 0.6500000 02 P= 0.0 G= 0.0  
 THE CURRENT VALUE OF THE PROB-MII.E.X(11) ) VARS.ARE;

X1= 0.20000000 01 X2= 0.20000000 01  
 THE CONSTRAINT VALUES  
 0.22073550 01 0.96027920 00 0.10000000 01

\*\*\*\*\*THE FEASIBLE STARTING POINT TO BE USED IS ...  
 F= 0.6500000 02 P= 0.0 G= 0.0  
 THE CURRENT VALUE OF THE PROB-MII.E.X(11) ) VARS.ARE;

X1= 0.20000000 01 X2= 0.20000000 01  
 THE CONSTRAINT VALUES  
 0.22073550 01 0.96027920 00 0.10000000 01

\*\*\*\*\*  
 POINT= 3 DOTT= 0.9541129D-05 RHO= 0.1000000 03 PHASE= 2  
 F= 0.4609753D 02 P= -0.3344122D 03 G= -0.3539025D 03  
 THE CURRENT VALUE OF THE PROB-MII.E.X(11) ) VARS.ARE;

X1= 0.37505537D 01 X2= 0.36536666D 01  
 THE CONSTRAINT VALUES  
 0.11163480 01 0.1670811D 01 0.1070327D 02

## LAGRANGE MULTIPLIERS

F= 0.4609753D 02 P= -0.3344122D 03 G= -0.3539025D 03  
 THE CURRENT VALUE OF THE PROB-MII.E.X(11) ) VARS.ARE;

X1= 0.37505537D 01 X2= 0.36536666D 01  
 THE CONSTRAINT VALUES  
 0.89577840 02 0.5985116D 02 0.9342937D 01

SENSITIVITY OF SOLUTION VECTOR WRT PROBLEM PARAMETERS  
 \*\*\*\*\*

PROBLEM PARAMETERS ARE  
 \*\*\*\*\*

C5 C6

THE VALUES OF SIG(I) IN DLPN ARE:  
 E 0.10000000 01 0.0

THE VALUES OF ALF RETURNED BY INVERS ARE :  
 F 0.60008081D-02 -0.22297426D-02

IN EVNED - THE VALES OF CVES XX(11) ARE :

(G) 6 0.460975260 02

THE VALS. OF DM(J) : I.E. PARTIALS OF NEW CVFS W.R.T. ORIGINAL PROB. VARS. (N BY 1) PREMULT. BY ALF ;

(H) 0.226782740-02 -0.462971150-02 0.135621830-01 -0.108014550 00

THE VALS. OF DIAGA (I,J) I.E. SECOND PARTIALS OF NEW CVFS W.R.T. Y & X PREMULT. BY ALF;

(I) 3 -0.222974260-02 0.0  
4 0.0 0.159997930-02  
5 0.0 0.0  
6 0.0 0.0

THE VALS OF DELEX(I,J) : PARTIALS OF NEW CVF WRT SENS VAR. Y;

(J) 3 0.365366660 01 0.0  
4 0.0 0.132190350 01  
5 0.0 0.0  
6 0.0 0.0

(K) THE VALUES OF DELA(J) ARE:  
0.465139590 00 -0.314990030 00

THE ESTIMATES OF ZEROITH ORDER SENSITIVITY OF XI WRT PROBLEM PARAMETERS ARE RESPECTIVELY :  
-----

-0.465140 00 0.314990 00

RESPECTIVE ELASTICITIES ARE :  
-----

-0.124020 00 -0.125980 00

THE VALUES OF SIG(I) IN DLPEN ARE:  
0.0 0.100000000 01

THE VALUES OF ALF RETURNED BY INVERS ARE :  
-0.222974260-02 0.850058150-02

IN EXDELP : THE VALS. OF CVFS XK(J) ARE ;  
1 0.375055370 01 2 0.365366660 01  
6 0.460975260 02

3 -.111634750 01 4 0.167081140 01 5 0.137032730 02

THE VALS. OF DM(J) : I.E. PARTIALS OF NEW CVFS W.R.T. ORIGINAL PROB. VARS. (N BY 1) PREMULT. BY ALF ;

0.682940270-02 0.939234690-02 0.237351510-01 0.401353670-01

THE VALS. OF DIAGA (I,J) I.E. SECOND PARTIALS OF NEW CVFS W.R.T. Y & X PREMULT. BY ALF;

3 0.850058150-02 0.0  
4 0.0 -0.594510250-03  
5 0.0 0.0  
6 0.0 0.0

THE VALS OF DELEX(I,J) : PARTIALS OF NEW CVF WRT SENS VAR. Y;

3 0.365366660 01 0.0  
4 0.0 0.132190350 01  
5 0.0 0.0  
6 0.0 0.0

THE VALUES OF DELA(J) ARE:  
0.276368840 01 0.480335270 00



## EXAMPLE-2 : SOLUTION AND SENSITIVITY ANALYSIS OF HUPKINS PROBLEM

SYNPUT PARAMETER CARD

N= 8 NN= 14 M= 12

M2= 0 ITEST= 1 IPR= 0 ITES= 0 NSWIT= 0 IID= 9

TYPE NAME	TYPE #	TYPE DESCRIP.	VAR. NAME	NUMBR. ASSND.	ARG. VARIAB. NAME	NUMBR.	FUNCT. TYPE	NUMBR. ASSND.	CONSTANT C1	CONSTANT C2
V		ORIG. VARIABLE	X1	1	0.180000 U3					
V		ORIG. VARIABLE	X2	2	0.740000 U3					
V		ORIG. VARIABLE	X3	3	0.268000 U4					
V		ORIG. VARIABLE	X4	4	0.146500 U5					
V		ORIG. VARIABLE	X5	5	0.400000 U0					
V		ORIG. VARIABLE	X6	6	0.230000 U1					
V		ORIG. VARIABLE	X7	7	0.600000 U0					
V		ORIG. VARIABLE	X8	8	0.200000 U0					
P		SENSITIVITY PAR	A1					1	0.240000 U2	
P		SENSITIVITY PAR	A2					2	0.120000 U2	
P		SENSITIVITY PAR	A3					3	0.600000 U0	
P		SENSITIVITY PAR	A4					4	0.400000 U0	
P		SENSITIVITY PAR	A5					5	0.100000 U0	
P		SENSITIVITY PAR	A6					6	0.150000 U0	
P		SENSITIVITY PAR	A7					7	-0.150000 U0	
P		SENSITIVITY PAR	A8					8	0.110000 U2	
P		SENSITIVITY PAR	A9					9	-0.240000 U2	
Q	4	QUADRATIC	X9	9	X4	4	LIN	2	0.0	0.0
Q	4	QUADRATIC	X9	9	X6	6	LIN	2	0.0	0.0
Q	4	QUADRATIC	X10	10	X3	3	LIN	2	0.0	0.0
Q	4	QUADRATIC	X10	10	X6	6	LIN	2	0.0	0.0
S	3	SEPARABLE	X11	11	X1	1	LIN	3	0.240000 U0 U2	0.0
S	3	SEPARABLE	X11	11	X2	2	LIN	3	0.120000 U0 U2	0.0
S	3	SEPARABLE	X11	11	X3	3	LIN	3	0.600000 U0 U0	0.0
S	3	SEPARABLE	X11	11	X4	4	LIN	3	0.400000 U0 U0	0.0
S	3	SEPARABLE	X11	11	X10	10	LIN	3	0.100000 U0 U0	0.0
S	3	SEPARABLE	X11	11	X9	9	LIN	3	0.150000 U0 U0	0.0
Q	4	QUADRATIC	X11	11	X9	9	LIN	3	-0.150000 U0 U0	0.0
Q	4	QUADRATIC	X11	11	X5	5	LIN	2	0.0	0.0
Q	4	QUADRATIC	X11	11	X1	1	LIN	3	0.110000 U0 U2	0.0
Q	4	QUADRATIC	X11	11	X7	7	LIN	2	0.0	0.0
Q	4	QUADRATIC	X11	11	X1	1	LIN	3	-0.240000 U0 U2	0.0
Q	4	QUADRATIC	X11	11	X8	8	LIN	2	0.0	0.0
S	3	SEPARABLE	X12	12	X1	1	LIN	3	0.122500 U0 U2	0.0
S	3	SEPARABLE	X12	12	X10	10	LIN	2	0.0	0.0
S	3	SEPARABLE	X12	12	X9	9	LIN	3	0.700000 U0 U0	0.0

北京市疾病预防控制中心

1	0.2740000	02
2	0.1200000	02
3	0.5000000	00
4	0.4000000	00
5	0.1000000	00
6	0.1500000	00
7	-0.1500000	00
8	0.1100000	02
9	-0.2240000	02
0	0.0	
1	0.1000000	01
2	0.1225000	02
3	0.7000000	00
4	-0.2000000	00
5	0.3000000	01
6	0.5000000	00
7	-0.1000000	00
8	-0.2000000	01
9	0.5000000	00
0	-0.2500000	02
1	0.5000000	00
2	-0.5000000	02
3	-0.1000000	02
4	0.1000000	00
5	0.5000000	00
6	-0.5000000	02
7	0.5000000	00
8	0.1000000	00
9	-0.1000000	01
0	-0.2000000	02
1	-0.1000000	01
2	0.2500000	05
3	0.3400000	05
4	0.5000000	00
5	0.3600000	00

CVF, CFF, VALUE OF EQUATION  
BOLD IN CELL(1)

11/11/11

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- 71 -

5 X1 1 0.0 -10  
 6 X2 2 0.0 -10  
 7 X3 3 0.0 -10  
 8 X4 4 0.0 -10  
 9 X5 5 0.0 -10  
 10 X6 6 0.0 -10  
 11 X7 7 0.0 -10  
 12 X8 8 0.0 -10

RELATIVE LOCATION: ITAB(.,.)  
 CVP# SEPARABLE QUADRATIC

1 1  
 10 1 2  
 11 1 3  
 12 7 6  
 13 10 7  
 14 11 7  
 15 18 10

SEPARABLE DEFINITIONS : IEP TAB(.,.)

ARG	ARG#	TRAN#	TRANSF	C1	LUC.IN	C2	LUC.IN	CC1
1	X1	1	3	LIN	0.24000 02	1	0.0	10
2	X2	2	3	LIN	0.12000 02	2	0.0	10
3	X3	3	3	LIN	0.00000 00	3	0.0	10
4	X4	4	3	LIN	0.00000 00	4	0.0	10
5	X10	10	3	LIN	0.10000 00	5	0.0	10
6	X9	9	3	LIN	0.15000 00	6	0.0	10
7	X1	1	3	LIN	0.12250 02	12	0.0	10
8	X10	10	2	IDN	0.0	10	0.0	10
9	X9	9	3	LIN	0.70000 00	13	0.0	10
10	X6	6	33	QDR	0.10000 01	11	0.3000 01	15
11	X1	1	11	PUM	0.50000 00	16	-0.1000 03	17
12	X2	2	3	LIN	-0.20000 01	13	0.0	10
13	X3	3	11	PUM	0.50000 00	19	-0.2500 02	20
14	X4	4	11	PUM	0.50000 00	21	-0.5000 02	22
15	X5	5	3	LIN	-0.10000 02	23	0.0	10
16	X13	13	3	LIN	0.10000 03	24	0.0	10
17	X8	8	3	LIN	0.10000 03	26	0.0	10

QUADRATIC DEFINITIONS : IADTB(.,.) : EACH LINE REFERS TO A PAIR OF ARGUMENTS

ARG	ARG#	TRAN#	TRANSF	APR	ARG#	TRAN#	TRANSF	C1	LUC.IN	C2	LUC.IN
1	X4	4	2	IDN	0.0	10	0.0	10	0.0	10	0.0
2	X3	3	2	LIN	0.0	10	0.0	10	0.0	10	0.0
3	X9	9	3	LIN	-0.15000 00	7	0.0	10	0.0	10	0.0
4	X1	1	3	LIN	0.11000 02	9	0.0	10	0.0	10	0.0
5	X1	1	3	LIN	-0.20000 02	9	0.0	10	0.0	10	0.0
6	X9	9	3	LIN	-0.20000 03	14	0.0	10	0.0	10	0.0
7	X1	1	11	PUM	0.50000 00	25	-0.3000 02	26	0.0	10	0.0
8	X1	1	3	LIN	-0.10000 02	29	0.0	10	0.0	10	0.0
9	X1	1	3	LIN	-0.20000 02	31	0.0	10	0.0	10	0.0

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[illegible]

D. LUBOVICH-02

F = -0.1041500 35 P = 0.0 G = 0.0  
THE CURRENT VALUE OF THE PROB-M(1.2.X(1) ) VARS.ALT;

[illegible]

DEL P VECTOR

0.501992680 01 -0.320701780 01 0.210614790 01 0.126376870 01 -0.000132700 04 0.113630100 03 -0.333749200 03  
0.103175050 03

0.0  
-0.123591230 04 -J.175165760 05 -0.6648043640 05 0.498175340 06 -0.267072022 01 -0.936937140 00 -0.100000000 01

[illegible]

```

POINT= 1      OUT= 0.5065309 06      KHI= 0.1100000 03      45041000= 0.1204932 05      PHASE= 1
F= 0.0        P= -0.25066010 04      W= -0.1200000 04      R21.94= -0.7500010 04      H= 0.0
THE CURRENT VALUE OF THE PROB-M(I,E,X(1) VARS,ARE;

```

THE CONSTRAINT VALUES

0.15616370	33	0.62392630	04	0.15749370	00	0.10000000	00	0.15369210	03	1.36760950	03
0.87483460	03	0.25267343	05	0.34259630	00	0.22789920	01	0.51371350	00	0.20330000	00

\*\*\*\*\*THE FEASIBLE STARTING POINT TO BE USED IS ...

$F = -0.10377410$  05  $P = -0.35086311$  14  $G = 0.0$   
THE CURRENT VALUE OF THE PROB-M(I,L,X(I)) VARS ARE;

$\alpha_1 = 0.153692049$  J3  
 $\alpha_6 = 0.227491811$  J1  
 THE COEFFICIENT VALUES

0.1561633	0.3	0.6259266	0.6	0.1574337	0.3	0.1000000	0.0	0.1536743	0.3	0.3070755	0.3
0.3740346	0.3	0.2567049	0.3	0.3525963	0.3	0.2274946	0.3	0.2717139	0.3	0.2730243	0.3

1568

$$\begin{aligned} & -0.1055(3\sigma I) \quad \sigma I = 0.5 \times (1.62 f_{10}) = 0.81 \\ & 0.1162(12\sigma I) \quad 12\sigma I = 0.5 \times (1.62 f_{10}) = 0.81 \end{aligned}$$
[illegible]



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POINT= 1 DUT= 0.1227300 03 MAG= 0.1000000 03 PHASE= 2  
F= -0.1103530 05 P= -0.1223500 05 MAG= -0.1223500 04 H= 0.0  
THE CURRENT VALUE OF THE PROB-M.I.E.X(1) VARS ARE:

X1= 0.1140040 03 X2= 0.3899910 01 X3= 0.8100000 03 X4= 0.2727170 03 X5= 0.4500510 00  
X6= 0.2414390 01 X7= 0.6977040 00 X8= 0.2700000 00  
THE CONSTRAINT VALUES  
0.3208640 05 0.9398930 04 0.1000000 01 0.1100000 03 0.4300440 03  
0.8109130 05 0.2727170 05 0.4500510 01 0.2414390 01 0.6977040 00 0.2700000 00

JEL P VECTOR  
-0.1345300 01 -0.1415370 01 0.3288010 00 -0.1163470 01 0.3057550 03 -0.3981310 03 -0.4609620 02  
0.25711260 03

SECOND ORDER MOVE VECTOR  
0.15391240 02 -0.1088910 04 0.1103650 05 -0.0702100 02 -0.3266030 01 -0.4251700 01  
0.15288610 01

POINT= 2 DUT= 0.2684270 04 MAG= 0.1000000 03 PHASE= 2  
F= -0.1160530 05 P= -0.1301790 05 MAG= -0.1280360 05 H= 0.0  
THE CURRENT VALUE OF THE PROB-M.I.E.X(1) VARS ARE:

X1= 0.1178860 03 X2= 0.1151270 03 X3= 0.1031430 04 X4= 0.3539870 05 X5= 0.4564690 00  
X6= 0.1585830 01 X7= 0.6869880 00 X8= 0.2034540 00  
THE CONSTRAINT VALUES  
0.9159090 03 0.3337480 04 0.4315060 01 0.9614560 01 0.1178360 03 0.1151270 03  
0.1089150 04 0.3539870 05 0.4564690 01 0.1585830 01 0.6869880 00 0.2034540 00

JEL P VECTOR  
0.4063820 01 0.15584320 01 0.43427560 00 0.10684000 03 -0.15000000 04 0.67279700 03 0.20347400 03  
-0.34063020 03

SECOND ORDER MOVE VECTOR  
-0.12024320 02 -0.49141490 02 0.11223050 04 0.3614470 04 0.30507840 04 -0.15912240 00 -0.10100060 01  
0.62556310 02

POINT= 3 DUT= 0.6333810 03 MAG= 0.1000000 03 PHASE= 2  
F= -0.1194350 05 P= -0.1523880 05 MAG= -0.1314350 05 H= 0.0  
THE CURRENT VALUE OF THE PROB-M.I.E.X(1) VARS ARE:

X1= 0.11145180 03 X2= 0.8816620 02 X3= 0.1690030 04 X4= 0.3133600 05 X5= 0.4568850 00  
X6= 0.1500630 01 X7= 0.6915810 00 X8= 0.2072030 00  
THE CONSTRAINT VALUES  
0.2115040 03 0.4001310 04 0.4311440 01 0.9279040 01 0.11145180 03 0.4568850 00  
0.1690030 04 0.3133600 05 0.4568850 01 0.1500630 01 0.6915810 00 0.2072030 00

ORTHOGONAL MOVE

JEL P VECTOR  
-0.54218270 01 -0.25477190 01 0.45711300 01 -0.9203710 01 0.1802940 04 -0.39415120 03 -0.27111220 02  
0.56966280 03

SECOND ORDER MOVE VECTOR  
-0.12057900 03 0.52600150 01 0.8360450 02 0.4273430 04 0.1199700 02 -0.11792320 00 0.1000000 01  
0.0

POINT= 4 DUT= 0.7553300 02 MAG= 0.1000000 03 PHASE= 2  
F= -0.1192420 05 P= -0.1523880 05 MAG= -0.1314350 05 H= 0.0  
THE CURRENT VALUE OF THE PROB-M.I.E.X(1) VARS ARE:

X1= 0.1093330 02 X2= 0.2121430 02 X3= 0.1711050 04 X4= 0.3740000 00 X5= 0.4568850 00

\*\*\*\*\*  
POINT= 24 DUE= 0.45570760-03  
F= -0.1257370 05 P= -0.1257370 05 G= -0.1257370 05 H= 0.0  
THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.11921590 03 X2= 0.08906590-02 X3= 0.47522150 04 X4= 0.39908110 05 X5= 0.49999920 00  
X6= 0.11335970 01 X7= 0.13701800 00 X8= 0.25999970 03  
THE CONSTRAINT VALUES  
0.3432590-01 0.4999990 00 0.5225990-03 0.1901460-04 0.1192220 03 0.1370180 00 0.2599990 00  
0.4759220 04 0.3990810 00 0.4999990 00 0.1192220 03 0.1370180 00 0.2599990 00  
APPARENTLY ROUND OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE VIOLATION OF THIS SUBPROBLEM.

2ND ORDER ESTIMATES  
F= -0.12573840 05 P= -0.12573720 05 G= -0.12573840 05 H= 0.0  
THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.11922950 04 X2= 0.89291720-04 X3= 0.47591900 04 X4= 0.39908940 05 X5= 0.49999990 00  
X6= 0.11335970 01 X7= 0.13701800 00 X8= 0.25999970 03  
THE CONSTRAINT VALUES  
0.3432590-01 0.4999990 00 0.5225990-03 0.1901460-04 0.1192220 03 0.1370180 00 0.2599990 00  
0.4759220 04 0.3990810 00 0.4999990 00 0.1192220 03 0.1370180 00 0.2599990 00

1ST ORDER ESTIMATES  
F= -0.12573840 05 P= -0.12573710 05 G= -0.12573840 05 H= 0.0  
THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.11922880 03 X2= 0.86840100-04 X3= 0.47591820 04 X4= 0.39909010 05 X5= 0.49999990 00  
X6= 0.11335970 01 X7= 0.13701800 00 X8= 0.25999970 03  
THE CONSTRAINT VALUES  
0.3432590-01 0.4999990 00 0.5225990-03 0.1901460-04 0.1192220 03 0.1370180 00 0.2599990 00  
0.4759220 04 0.3990810 00 0.4999990 00 0.1192220 03 0.1370180 00 0.2599990 00

LAGRANGE MULTIPLIERS  
F= -0.12573840 05 P= -0.12573710 05 G= -0.12573840 05 H= 0.0  
THE CURRENT VALUE OF THE PROB-M(I.E.X(I)) VARS.ARE;

X1= 0.11922880 03 X2= 0.86840100-04 X3= 0.47591820 04 X4= 0.39909010 05 X5= 0.49999990 00  
X6= 0.11335970 01 X7= 0.13701800 00 X8= 0.25999970 03  
THE CONSTRAINT VALUES  
0.3432590-01 0.4999990 00 0.5225990-03 0.1901460-04 0.1192220 03 0.1370180 00 0.2599990 00  
0.4759220 04 0.3990810 00 0.4999990 00 0.1192220 03 0.1370180 00 0.2599990 00

SENSITIVITY OF SOLUTION VECTOR WRT PROBLEM PARAMETERS  
\*\*\*\*\*  
PROBLEM PARAMETERS ARE  
\*\*\*\*\*

A1	A2	A3	A4	A5	A6	A7	A8
0.11922880 03	0.69973150-02	0.47590970 04	0.39908940 05	0.49999990 00	0.1192220 03	0.1370180 00	0.2599990 00
0.2599990 00	0.4759220 04	0.3990810 00	0.4999990 00	0.1192220 03	0.1370180 00	0.2599990 00	0.4759220 04

DEL P VECTOR  
0.46901550 02 -0.26999990-03 -0.41707750-03 -0.45169910 04 -0.45169910 04 -0.45169910 04 -0.45169910 04 -0.45169910 04

DEL Q VECTOR  
0.46901550 02 -0.26999990-03 -0.41707750-03 -0.45169910 04 -0.45169910 04 -0.45169910 04 -0.45169910 04 -0.45169910 04

THE ESTIMATES OF ABSOLUTE SENSITIVITY OF AT WITH PROBLEM PARAMETERS ARE \*\*\*\*\*

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FROM COPY NUMBERED 100

[illegible]



THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X4 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.104030 05 -0.590950 00 -0.412370 06 -0.364220 07 -0.666620 08 -0.412170 09 -0.206350 10 -0.191290 04 -0.339220 06

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X4 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.114940 05 -0.636140 00 -0.456100 06 -0.400970 07 -0.516190 08 -0.453120 09 -0.226960 10 -0.211790 04 -0.342190 04

RESPECTIVE ELASTICITIES ARE :

=====

-0.685790 01 -0.191290 03 -0.635710 01 -0.401790 02 -0.129340 01 -0.170590 02 0.352690 01 -0.583620 00 0.207790 01

DEL P VECTOR

0.119236980 03

0.299985610 00

SECOND ORDER MOVE VECTOR

0.692087460 03

-0.419223380 14

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X5 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.301160 06 -0.189650 10 -0.106740 04 -0.113160 03 -0.120740 04 -0.124100 03 -0.109990 03 -0.509970 07 -0.9093450 07

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X5 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.828070 06 -0.280530 09 -0.333490 04 -0.274490 03 -0.378130 04 -0.311230 03 -0.145240 03 -0.194290 06 -0.246350 06

RESPECTIVE ELASTICITIES ARE :

=====

-0.397480 04 -0.673280 08 -0.400190 04 -0.219000 03 -0.756240 03 -0.953710 04 0.447740 04 -0.339360 05 0.117240 04

DEL P VECTOR

0.119236980 03

0.299985610 00

SECOND ORDER MOVE VECTOR

0.231385520 01

-0.193694250 09

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X6 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

0.126990 00 0.685170 05 0.544430 01 0.449470 02 0.614540 01 1.507310 02 0.262650 02 0.222940 01 0.230950 01

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X6 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

0.139210 00 0.715970 05 0.593100 01 0.499240 02 0.609770 01 0.351440 02 0.212820 02 0.245750 01 0.417610 01

RESPECTIVE ELASTICITIES ARE :

=====

0.244720 01 0.157490 04 0.413960 01 0.173020 02 0.590470 03 0.122560 01 -0.366070 04 1.233990 00 -0.064120 00



- 77 -

DEL P VECTOR  
 0.41923690-03 0.694731650-02 0.475994730-04 0.349045400-03 0.499994100-00 0.113302450-01 0.180692913-00  
 0.299985610-00

SECOND ORDER MOVE VECTOR  
 -0.70970070-01 -0.09577170-07 -0.38112460-01 -0.445775200-00 0.13991591-09 0.92191289-04 0.135669940-02  
 -0.702828230-08

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF AT RT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.155400 01 -0.909070-04 -0.028345 02 -0.020650 03 -0.713010 02 -0.597800 01 -0.290920 03 -0.324530 00 -0.466017 00

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF AT RT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.175010 01 -0.113040-03 -0.706090 02 -0.092560 03 -0.401710 02 -0.672230 03 -0.336110 03 -0.365160 00 -0.525510 00

RESPECTIVE ELASTICITIES ARE :

=====

-0.224520 03 -0.725090-02 -0.220650 03 -0.126700 04 -0.423530 02 -0.533990 03 0.209990 03 -0.219710 02 0.673520 02

DEL P VECTOR

0.119236980-03 0.694731650-02 0.475994730-04 0.349045400-03 0.499994100-00 0.113302450-01 0.180692913-00  
 0.299985610-00

SECOND ORDER MOVE VECTOR

0.483348970-05 -0.120745990-10 -0.813179110-05 -0.662254230-04 -0.419223380-14 -0.13697250-04 -0.702428230-08  
 0.155032610-07

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF KB RT PROBLEM PARAMETERS ARE RESPECTIVELY :

0.737790-04 0.434640-08 0.300450-02 0.251940-01 0.340620-02 0.285630-01 0.162310-01 0.140670-04 0.215940-04

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF KB RT PROBLEM PARAMETERS ARE RESPECTIVELY :

0.924230-04 0.961130-08 0.367310-02 0.303010-01 0.416910-02 0.349180-01 0.174590-01 0.172040-04 0.273970-04

RESPECTIVE ELASTICITIES ARE :

=====

0.739420-02 0.384490-06 0.734600-02 0.410700-01 0.136310-02 0.174630-04 -0.372970-02 0.009990-03 -0.222190-02

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EXAMPLE-3 : SOLUTION AND SENSITIVITY ANALYSIS OF CORRUGATED BULKHEAD									
SYMPUT PARAMETER CARD									
N=	6	NN=	45	M=	19	MZ=	0	ITEST=	1
						IPR=	0	ITES=	0
						NSMII=	0	IID=	15
TYPE	TYPE	TYPE	VAR.	NUMBR.	ARG.	NUMBR.	FUNCT.	CONSTANT	CONSTANT
NAME	#	DESCRIP.	NAME	ASSND.	VARIABLE	ASSNC.	TYPE	C1	C2
V	ORIG.	VARIABLE	X1	1	0.458000	02			
V	ORIG.	VARIABLE	X2	2	0.432000	02			
V	ORIG.	VARIABLE	X3	3	0.305000	02			
V	ORIG.	VARIABLE	X4	4	0.120000	01			
V	ORIG.	VARIABLE	X5	5	0.120000	01			
V	ORIG.	VARIABLE	X6	6	0.130000	01			
P	SENSITIVITY	PAR	LT					1	0.495000
P	SENSITIVITY	PAR	LM					2	0.385000
P	SENSITIVITY	PAR	LB					3	0.315000
P	SENSITIVITY	PAR	TTM					4	0.105000
P	SENSITIVITY	PAR	TMM					5	0.105000
P	SENSITIVITY	PAR	TBM					6	0.105000
P	SENSITIVITY	PAR	K2					7	0.150000
D	DATA	DEFINITION	E						0.800000000 00
D	DATA	DEFINITION	GB						0.37366000 04
D	DATA	DEFINITION	KHT						0.345600000-04
D	DATA	DEFINITION	KHM						0.651000000-04
D	DATA	DEFINITION	KHB						0.893300000-04
D	DATA	DEFINITION	CT						0.11170000-01
D	DATA	DEFINITION	CM						0.137650000-01
D	DATA	DEFINITION	CB						0.155660000-01
S	3	SEPARABL	Z1	7				10	0.200000000 01
S	3	SEPARABL	Z1	7				11	0.200000000 01
S	3	SEPARABL	Z2	8					-0.100000000 01
S	3	SEPARABL	Z2	8					0.0
Q	4	QUADRATC	Z3	9				10	0.500000000 00
Q	4	QUADRATC	Z3	9				11	0.500000000 00
Q	4	QUADRATC	Z4	10				12	0.0
Q	4	QUADRATC	Z4	10				13	0.500000000 00
Q	4	QUADRATC	Z5	11				14	0.0
Q	4	QUADRATC	Z5	11				15	0.0
S	3	SEPARABL	Z6	12				16	0.0
S	3	SEPARABL	Z6	12				17	0.0
Q	4	QUADRATC	Z7	13				18	0.0
Q	4	QUADRATC	Z7	13				19	0.0
Q	4	QUADRATC	Z8	14				20	0.0
Q	4	QUADRATC	Z8	14				21	0.0
Q	4	QUADRATC	Z9	15				22	0.0
Q	4	QUADRATC	Z9	15				23	0.0
S	3	SEPARABL	Z10	16				24	0.0
S	3	SEPARABL	Z10	16				25	0.0
S	3	SEPARABL	Z11	17				26	0.0
S	3	SEPARABL	Z11	17				27	0.0
S	3	SEPARABL	Z12	18				28	0.0
S	3	SEPARABL	Z12	18				29	0.0
S	3	SEPARABL	Z13	19				30	0.0
S	3	SEPARABL	Z13	19				31	0.0
S	3	SEPARABL	Z14	20				32	0.0
S	3	SEPARABL	Z14	20				33	0.0
S	3	SEPARABL	Z15	21				34	0.0
S	3	SEPARABL	Z15	21				35	0.0





- 80 -

NONLINEAR PROGRAMMING ROUTINE-304, VERSION 4 04/10/71

N= 6 4= 19 12= 0  
MAX. TIME= 0.4000000 03 K= 0.1000000 03 1210= 0.0000000 01 1211= 0.10000000-04 1212= 0.10000000-05

OPTIONS SELECTED

3 2 1 1 1 1 2 1 1 1

TOLERANCES

0.10000000-02 0.10000000-02

SECOND SET OF OPTIONS

1

F= 0.6179000 07 P= 0.0 Q= 0.0  
THE CURRENT VALUE OF THE PROB-MIL.E.X(1) 1 VARS.ARE;

X1= 0.45000000 02	X2= 0.43200000 02	X3= 0.30500000 02	X4= 0.12000000 01	X5= 0.12000000 01
X6= 0.13000000 01				
THE CONSTRAINT VALUES				
0.45000000 02	0.43200000 02	0.30500000 02	0.12700000 02	0.23712310 03
0.33430000 03	0.19352510 04	-0.40607190 03	0.23356720 04	0.15030000 00
0.56715360 00	0.15000000 00	0.41956300 00	0.45535230 00	0.25300000 00
0.47754880 00				0.43707720 00

ORTHOGONAL MOVE

DEL P VECTOR

-0.31385370 02 -0.29407000 03 0.13432742 04 0.19132730 04 0.13589573 05 0.15783784 04

SECOND ORDER MOVE VECTOR

0.79081510 01 0.19889040 01 0.10000000 01 0.0 0.0 0.0

\*\*\*\*\*

POINT= 1 DUFF= 0.51019610 03 KMF= 0.10000000 07 MAGNITUDE= 0.13687040 05 PHASE= 1  
F= 0.0 P= -0.41337020 04 Q= -0.19000000 04 KSGMA= -0.41347020 04 H= 0.0  
THE CURRENT VALUE OF THE PROB-MIL.E.X(1) 1 VARS.ARE;

X1= 0.33704150 02	X2= 0.45188904 02	X3= 0.21500000 02	X4= 0.12000000 01	X5= 0.12000000 01
X6= 0.13000000 01				
THE CONSTRAINT VALUES				
0.53704150 02	0.45188900 02	0.31500000 02	0.13688900 02	0.36753450 03
0.42448010 03	0.28351640 04	0.84734400 02	0.33566560 04	0.15030000 00
0.54492300 00	0.15000000 00	0.31370730 00	0.42797470 00	0.25000000 00
0.44658950 00				0.31397890 00

\*\*\*\*\*THE FEASIBLE STARTING POINT TO BE USED IS ...

F= 0.62892580 07 P= -0.41337020 04 Q= 0.0  
THE CURRENT VALUE OF THE PROB-MIL.E.X(1) 1 VARS.ARE;

X1= 0.53704150 02	X2= 0.45188904 02	X3= 0.31500000 02	X4= 0.12000000 01	X5= 0.12000000 01
X6= 0.13000000 01				
THE CONSTRAINT VALUES				
0.53704150 02	0.45188900 02	0.31500000 02	0.13688900 02	0.36753450 03
0.42448010 03	0.28351640 04	0.84734400 02	0.33566560 04	0.15030000 00
0.54492300 00	0.15000000 00	0.31370730 00	0.42797470 00	0.25000000 00
0.44658950 00				0.31397890 00

ORTHOGONAL MOVE

DEL P VECTOR

0.39097940 04 0.37779400 03 -0.30830720 05 -0.41220000 04 -0.30830720 04 -0.15030000 01



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\*\*\*\*\*  
 POINT= 1  
 F= 0.2924760 07  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(1)) VARS.ARE:  
 X1= 0.52350353 02  
 X2= 0.44841570 02  
 X3= 0.31447079 02  
 X4= 0.10540525 01  
 X5= 0.12000000 01  
 THE CONSTRAINT VALUES  
 U.52350550 02  
 0.41380710 03  
 0.39885810 00  
 0.45199610 00  
 RMU= 0.10000000 03  
 G= 0.54843570 07  
 MAGNITUDE= 0.30005340 07  
 RSLUMA= -0.3143330 04  
 H= 0.0  
 PHASE= 2

\*\*\*\*\*  
 POINT= 2  
 F= 0.57933220 07  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(1)) VARS.ARE:  
 X1= 0.53330010 02  
 X2= 0.47661965 02  
 X3= 0.33867900 02  
 X4= 0.10500516 01  
 X5= 0.10500775 01  
 THE CONSTRAINT VALUES  
 0.53330010 02  
 0.47661970 02  
 0.30223350 04  
 0.36733380 00  
 0.46909330 00  
 RMU= 0.10000000 03  
 G= 0.57914220 07  
 MAGNITUDE= 0.21455210 07  
 RSLUMA= -0.25374070 04  
 H= 0.0  
 PHASE= 2

## ORTHOGONAL MOVE

## DEL P VECTOR

0.939105310 04 0.379133390 05 -0.687793270 05 -0.216564250 06 -0.164730620 07 -0.135520450 07

## SECOND ORDER MOVE VECTOR

0.653339390 01 0.188123600 02 0.161476830 02 -0.445143470 05 -0.100000000 01 0.0

\*\*\*\*\*

\*\*\*\*\*  
 POINT= 3  
 F= 0.54846190 07  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(1)) VARS.ARE:  
 X1= 0.52216326 02  
 X2= 0.48098175 02  
 X3= 0.34472320 02  
 X4= 0.10500513 01  
 X5= 0.10500757 01  
 THE CONSTRAINT VALUES  
 0.52216830 02  
 0.48098170 02  
 0.28567930 03  
 0.36245820 00  
 0.15138500 00  
 RMU= 0.10000000 03  
 G= 0.54824150 07  
 MAGNITUDE= 0.14351140 07  
 RSLUMA= -0.15349420 04  
 H= 0.0  
 PHASE= 2

\*\*\*\*\*  
 POINT= 4  
 F= 0.54846190 07  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(1)) VARS.ARE:  
 X1= 0.52216326 02  
 X2= 0.48098175 02  
 X3= 0.34472320 02  
 X4= 0.10500513 01  
 X5= 0.10500757 01  
 THE CONSTRAINT VALUES  
 0.52216830 02  
 0.48098170 02  
 0.28567930 03  
 0.36245820 00  
 0.15138500 00  
 RMU= 0.10000000 03  
 G= 0.54824150 07  
 MAGNITUDE= 0.14351140 07  
 RSLUMA= -0.15349420 04  
 H= 0.0  
 PHASE= 2

## ORTHOGONAL MOVE

## DEL P VECTOR

0.931062490 04 0.372880300 05 -0.667137300 05 -0.211445920 06 -0.374362190 06 -0.136702860 07

## SECOND ORDER MOVE VECTOR

-0.445418430 01 0.174540480 01 0.241814950 01 -0.138947920 05 -0.713514360 05 -0.100000000 01

\*\*\*\*\*

\*\*\*\*\*  
 POINT= 5  
 F= 0.54846190 07  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(1)) VARS.ARE:  
 X1= 0.52216326 02  
 X2= 0.48098175 02  
 X3= 0.34472320 02  
 X4= 0.10500513 01  
 X5= 0.10500757 01  
 THE CONSTRAINT VALUES  
 0.52216830 02  
 0.48098170 02  
 0.28567930 03  
 0.36245820 00  
 0.15138500 00  
 RMU= 0.10000000 03  
 G= 0.54824150 07  
 MAGNITUDE= 0.14351140 07  
 RSLUMA= -0.15349420 04  
 H= 0.0  
 PHASE= 2

\*\*\*\*\*  
 POINT= 6  
 F= 0.54846190 07  
 THE CURRENT VALUE OF THE PROB-M(I.E.X(1)) VARS.ARE:  
 X1= 0.52216326 02  
 X2= 0.48098175 02  
 X3= 0.34472320 02  
 X4= 0.10500513 01  
 X5= 0.10500757 01  
 THE CONSTRAINT VALUES  
 0.52216830 02  
 0.48098170 02  
 0.28567930 03  
 0.36245820 00  
 0.15138500 00  
 RMU= 0.10000000 03  
 G= 0.54824150 07  
 MAGNITUDE= 0.14351140 07  
 RSLUMA= -0.15349420 04  
 H= 0.0  
 PHASE= 2

## ORTHOGONAL MOVE

## DEL P VECTOR

0.925815640 04 0.369110250 05 -0.659465060 05 -0.211563970 06 -0.334347990 06 -0.141462910 06

## SECOND ORDER MOVE VECTOR

-0.778216310 00 -0.120361250 01 -0.107000000 01 0.0 0.0 0.0

```
*****
PUNIT= 33      DUNIT= 0.14733570-02      RFG= 0.9165020-01      RALPH= 0.1363330 00      RMSE= 2
F= 0.52480230 01  P= 0.52480270 01  Q= 0.52430210 01  C01,PA= 0.4261,220 01  H= 0.0
WE CURRENT VALUE OF THE PROBABILITY OF A SUCCESS:
*****
```

[illegible]

APPARENTLY ROUND-OFF ERRORS PREVENT A MORE ACCURATE DETERMINATION OF THE MINIMUM OF THIS SUBPROBLEM.

## 2ND ORDER ESTIMATES

$F = 0.52480220$  OF  
 THE CURRENT VALUE OF  
 $\rho = 0.52480220$  OF  
 $G = 0.52480220$  OF  
 $K_{S1,MA} = 0.0$   
 $n = 0.0$

THE CONSTRAINT VALUES					
X1=	X2=	X3=	X4=	X5=	
0.578183220 02	0.578183220 02	0.559332970 02	0.221313330 02	0.55222720 03	0.2393400 03
0.105000000 01			0.233092330 04	0.11546313-08	0.22576460 00
			0.104132810 01	0.11795370-07	0.75112200-08
0.57818320 02	0.57818320 02	0.55933300 02			
0.34153520 03	0.35012100 04	0.87647100-04			
0.25376460 00	0.23255240-08	0.10413040 00			
0.30338920-08					

# 1ST ORDER ESTIMATES

F = 0.52480220 07  
THE CURRENT VALUE OF THE PROC-M(I,E,X(I)) VARS ARE:  
P = 0.52480220 07 G = 0.52480220 07 KSLGWA = 0.0 H = 0.0

THE CONSTRAINT VALUES						
X1=	X2=	X3=	X4=	X5=		
0.578183220 02	0.578183220 02	0.356832970 02	0.105000000 01	0.105000000 01		
X6=	0.105000000 01					
THE CONSTRAINT VALUES						
U.57818320 02	U.57818320 02	U.35683300 02	U.22133000 02	U.35292720 03	U.23083090 03	
0.31153520 03	0.35012170 04	0.77644430-04	U.22309200 04	U.10328810-08	U.25376450 00	
0.25376460 03	U.20640540-08	U.10413080 00	U.10413090 00	U.10527400-07	U.66589670-08	
U.26895380-08						

## LAGRANGE MULTIPLIERS

```

F = 0.52480220 DT      P = 0.52480240 DT      RSIGMA = 0.0
THE CURRENT VALUE OF THE PGM-M(I.E.X(1)) VARS.ARE:

```

THE CONSTRAINT VALUES					
X1=	X2=	X3=	X4=	X5=	X6=
0.578183220 02	0.578183220 02	0.355852570 02	0.491233000 02	0.271703500 03	0.722797190 03
0.105000000 01			0.413050000 01	0.20972000 01	0.308430020 00
			0.931225000 00	0.10030750 00	0.31750520 00
0.166901500 02	0.158901300 02	0.273503000 02			
0.313407100 03	0.278921100 04	0.271872000 02			
0.30483000 00	0.19255000 01	0.93324500 00			
0.1850427 06					

### SENSITIVITY OF SOLUTION VECTOR WITH RESPECT TO PARAMETERS

## PROBLEM 4 PARAMETERS ARE

$\Gamma$	L <sup>B</sup>	L <sup>C</sup>	L <sup>D</sup>	L <sup>E</sup>	K <sup>F</sup>
0.57618349 02	0.57618349 02	0.57618349 02	0.57618349 02	0.57618349 02	0.57618349 02

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0.125706020-07 0.900249740-06 0.167239250-08 0.142723130-17 -0.929278170-17 0.140122930-09

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

0.145720-06 0.124830-05 -0.449330-06 0.612560-04 -0.599430-04 0.572890-02 -0.572890-02

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X1 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.255490-07 -0.376380-06 0.172000-07 -0.253500-04 0.486330-04 0.540540-02 -0.540610-02

RESPECTIVE ELASTICITIES ARE :  
=====

-0.218740-06 -0.250620-05 0.420590-06 -0.478520-06 0.483560-06 0.981640-00 -0.140250-00

DEL P VECTOR

0.578183340-02 0.578183340-02 0.356883330-02 0.105000000-01 0.105000000-01 0.105000000-01 0.105000000-01

SECOND ORDER MOVE VECTOR

0.900249740-08 0.953874730-08 0.173435400-04 0.134015490-17 -0.861938310-17 0.140132980-09

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

0.136830-06 0.122840-05 -0.449270-06 0.575110-04 -0.948950-04 0.572890-02 -0.572890-02

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X2 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.239940-07 -0.355040-06 0.787570-07 -0.247440-04 0.464250-04 0.540540-02 -0.540610-02

RESPECTIVE ELASTICITIES ARE :  
=====

-0.205420-06 -0.236410-05 0.429070-06 -0.449360-06 0.443100-06 0.981640-00 -0.140250-00

DEL P VECTOR

0.578183340-02 0.578183340-02 0.356883330-02 0.105000000-01 0.105000000-01 0.105000000-01 0.105000000-01

SECOND ORDER MOVE VECTOR

0.167239250-08 0.173435400-08 0.337523720-09 0.223355200-13 -0.119235000-11 0.254464070-10

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X3 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

0.224980-07 0.933420-01 -0.839320-07 0.945750-05 -0.128330-12 0.104030-02 -0.104020-02

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X3 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.346690-04 0.889920-01 0.147110-07 -0.440500-05 -0.121120-12 0.981640-00 -0.981640-00

RESPECTIVE ELASTICITIES ARE :  
=====

-0.247440-07 0.953874730-08 0.173435400-04 -0.134015490-17 -0.861938310-17 0.140132980-09

DEL P VECTOR

0.578183340-02 0.578183340-02 0.356883330-02 0.105000000-01 0.105000000-01 0.105000000-01 0.105000000-01



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0.15272370-17 0.13918360-15 0.22035500-15 0.30775000-15 0.38105000-15 0.45421000-15

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X9 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.866950-10 -0.309900-10 -0.027700-10 0.041000 00 0.415200-08 0.812700-05 -0.419190-03

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X9 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

0.152020-10 0.130950-10 0.390830-15 0.841490 00 -0.180210-08 -0.350990-03 0.350120-03

RESPECTIVE ELASTICITIES ARE :

=====

0.716680-03 0.440150-08 0.103450-12 0.041490 00 -0.180210-08 -0.350990-03 0.350120-03

DEL P VECTOR

0.57818340 02 0.57818340 02 0.39083000 02 0.10500000 01 0.10500000 01 0.10500000 01

SECOND ORDER MOVE VECTOR

-0.92927870-17 -0.88193810-17 -0.119235080-11 0.390830000-21 0.323727000-13 -0.129052330-18

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X5 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

0.397810-10 -0.436090-10 0.397730-10 0.167240-07 0.091700 00 -0.050070-07 0.036460-07

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X5 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.695430-11 -0.120300-10 -0.694900-11 -0.714790-08 0.041490 00 0.259580-07 -0.304470-07

RESPECTIVE ELASTICITIES ARE :

=====

-0.327850-08 -0.441080-08 -0.208170-06 -0.718790-08 0.041490 00 0.259580-07 -0.304470-07

DEL P VECTOR

0.57818340 02 0.57818340 02 0.39083000 02 0.10500000 01 0.10500000 01 0.10500000 01

SECOND ORDER MOVE VECTOR

0.190132930-09 0.190132930-09 0.254468070-10 0.199421840-19 -0.129052330-18 0.218131080-11

THE ESTIMATES OF ZEROth ORDER SENSITIVITY OF X6 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

0.203610-08 0.181510-07 -0.708720-08 0.39083000-05 -0.149510-05 0.851700 00 0.641780-05

THE ESTIMATES OF FIRST ORDER SENSITIVITY OF X6 WRT PROBLEM PARAMETERS ARE RESPECTIVELY :

-0.3357100-09 -0.523400-03 0.129230-03 -0.363630-06 0.032900-06 0.041490 00 -0.242110-05

RESPECTIVE ELASTICITIES ARE :

=====

-0.164350-16 -0.191910-05 0.332700-06 -0.100220-06 0.032900-06 0.041490 00 -0.242110-05



APPENDIX 2

"Sensitivity Analysis in Nonlinear  
Programming Using Factorable  
Symbolic Input"  
(borrowed directly from [4])

THE GEORGE WASHINGTON UNIVERSITY  
School of Engineering and Applied Science  
Institute for Management Science and Engineering

SENSITIVITY ANALYSIS IN NONLINEAR PROGRAMMING  
USING FACTORABLE SYMBOLIC INPUT

by

Anura de Silva  
Garth P. McCormick

1. Introduction

This paper develops a methodology to utilize the input to nonlinear programs in symbolic factorable form to perform first order sensitivity analysis on the solution vector  $x(y) \in E^n$  of a general nonlinear program, where  $y \in E^k$  is the parametric vector. In Section 2 we discuss the key results of sensitivity analysis theory in nonlinear programming. In Section 3 we develop the formulae underlying the new methodology to utilize the symbolic factorable input to perform sensitivity analysis on the solution vector  $x(y)$ .

The above are implemented via the SUMT algorithm [4] for nonlinear programming which uses a penalty function technique. Section 4 briefly discusses the penalty function-based implementation of the sensitivity analysis methodology. An extrapolation result based on the penalty function parameter  $r$  is stated and incorporated in the computer code.

2. Relevant Results from NLP Sensitivity Theory

Consider the problem

$$\left. \begin{array}{l} \text{MIN} \quad F(x,y) \\ x \in E^n \\ \text{s.t.} \quad G(x,y) \geq 0 \\ \quad \quad H(x,y) = 0 \\ \text{where } G : E^n \times E^k \rightarrow E^m \\ \quad \quad H : E^n \times E^k \rightarrow E^p \end{array} \right\} \text{Problem } M(y)$$

By sensitivity information we mean the rates of change of the (locally) optimal objective function value and the corresponding optimizing vector with respect to changes in the parametric vector. If  $x(y)$  is an  $n$ -vector and  $y$  a  $k$ -vector, the first order sensitivity information for Problem  $M(y)$  is:

- (a) The rate of change of the solution value with respect to the parametric vector,  $D_y F[x(y), y] \in E^k$ .
- (b) The rate of change of the optimizing vector with respect to the parametric vector,  $\frac{dx(y)}{dy}$ , where

$$\frac{dx(y)}{dy} = \begin{bmatrix} \frac{\partial x_1(y)}{\partial y_1} & \dots & \frac{\partial x_1(y)}{\partial y_k} \\ \vdots & & \vdots \\ \frac{\partial x_n(y)}{\partial y_1} & \dots & \frac{\partial x_n(y)}{\partial y_k} \end{bmatrix} \quad (1)$$

The first order sensitivity results for the general nonlinear programming problem were first stated by Fiacco and McCormick in [3], Theorem 6, where they give conditions under which a smooth (once continuously differentiable, abbreviated OCD) trajectory of the optimizing Kuhn-Tucker triple  $[x(y), u(y), w(y)]$  exist for  $y$  close to an initial value  $y_0$ . The  $u$  and  $w$  are the Lagrange multiplier vectors associated with the inequality

constraints  $G$  and equality constraints  $H$  of Problem  $M(y)$ , in the Lagrangian,

$$L(x, y, u, w) = F(x, y) - \sum_{i=1}^m u_i G_i(x, y) + \sum_{j=1}^p w_j H_j(x, y) \quad (2)$$

where  $u_i, G_i; i = 1, \dots, m$  and  $w_j, H_j; j = 1, \dots, p$  are the components of the mappings  $G$  and  $H$  and the Lagrange multipliers  $u$  and  $w$ . Here,  $u(y)$  and  $w(y)$  are the values of these multipliers associated with the optimal value  $x(y)$ . They assume the following to hold; (a) twice continuous differentiability (abbreviated TCD) of  $L(x, y, u, w)$  in  $(x, y)$  in a neighborhood of  $(x(y_0), y_0)$ , (b) second order sufficiency conditions (abbrev. SOS) at  $(x(y_0), y_0)$ , (c) linear independence of binding constraints (abbrev. LIB) at  $(x(y_0), y_0)$ , (d) strict complementary slackness (abbrev. SCS). Here, (c) and (d) are regularity conditions they invoke to obtain the key result in a useful form. In a more recent paper Fiacco [2] obtains a similar result for somewhat more general parametric problem, such as  $M(y)$ .

The crux of the theory is the application of the implicit function theorem (see [6] for a general treatment of the implicit function theorem) to the first order Kuhn-Tucker system of  $M(y)$ , which is

$$\begin{aligned} \nabla_x L(x, y, u, w) &= 0 \\ u_i G_i(x, y) &= 0, \quad i = 1, \dots, m \\ H_j(x, y) &= 0, \quad j = 1, \dots, p. \end{aligned} \quad (3)$$

Consider a system of equations

$$D(x, y) = 0, \quad x \in E^n \quad \text{and} \quad y \in E^k \quad \text{Problem } P(y),$$

where  $D : E^n \times E^k \rightarrow E^\ell$ ,  $\ell$  being the number of scalar valued functions in  $D$ , which are assumed OCD. This ensures the existence of the Jacobians of  $D(x, y)$  with respect to  $x$  and  $y$ , denoted  $J_x D$  and  $J_y D$ , respectively. With the sensitivity analysis context in view, the implicit function theorem could be stated as follows:



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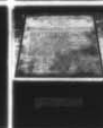
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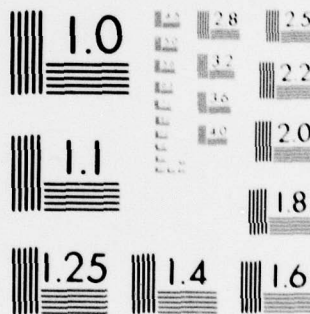
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Theorem 1. Suppose  $D : E^n \times E^k \rightarrow E^l$ ;  $(x(y_0), y_0)$  is a zero of  $D(x, y)$ ;  $D(x, y)$  is OCD near  $(x(y_0), y_0)$ ;  $J_x D$  is nonsingular at  $(x(y_0), y_0)$ .

Then, there exist open sets  $S_x(x(y_0)) \subset E^n$  and  $S_y(y_0) \subset E^k$  such that for any  $y \in \bar{S}_y(y_0)$  there exists a unique solution for  $x$  in terms of  $y$ , i.e.,  $x = x(y) \in S_x(x(y_0))$ , which is continuous in  $y$ . Moreover  $x(y)$  is OCD at  $y_0$  and

$$\nabla_y x(y_0) = - \left[ J_x D_0 \right]^{-1} \left[ J_y D_0 \right],$$

where the subscript 0 on  $D$  denotes evaluation at  $(x(y_0), y_0)$ .

If  $P(y)$  is compared with (3) the role of Theorem 1 in the context of sensitivity analysis of the solution vector becomes evident. The assumed TCD of the problem functions gives the necessary OCD for the system in (3). The additional conditions of linear independence of the binding constraint gradients and SCS were assumed by Fiacco and McCormick [3] and Fiacco [2] to produce desirable results such as the uniqueness of the Lagrange multipliers  $[u(y), w(y)]$  associated with the optimal  $x(y)$  and the invariance of the set of inequality constraints binding at  $x(y_0)$  in a neighborhood of  $y_0$ . The following representation for the Kuhn-Tucker triple,

$$X \triangleq (x, u, w)$$

$$\text{and } X(y) \triangleq [x(y), u(y), w(y)]$$

will be adopted for notational convenience. For the same reason, the first order Kuhn-Tucker system of (3) will be denoted as follows.

$$R(x, u, w, y) = 0$$

or

$$\bar{R}(x, y) = 0.$$

(3b)

Differentiating (3b) partially with respect to  $X$  yields the Jacobian of (3b) or (3) with respect to  $X$ ,

$$J_X \bar{R}(X, y) = \begin{bmatrix} \nabla_{xx}^2 L & -\nabla_x G_1, \dots, -\nabla_x G_m & \nabla_x H_1, \dots, \nabla_x H_p \\ u_1 \nabla_x^T G_1 & G_1 & 0 \\ \vdots & \ddots & \vdots \\ u_m \nabla_x^T G_m & 0 & G_m \\ \nabla_x^T H_1 & \vdots & \vdots \\ \vdots & 0 & \vdots \\ \nabla_x^T H_p & \vdots & \vdots \end{bmatrix} \quad (4)$$

Differentiating (3b) partially with respect to  $y$  yields the Jacobian with respect to this parametric vector,

$$J_y \bar{R}(X, y) = \left[ \nabla_{yx}^2 L^T, u_1 \nabla_y G_1, \dots, u_m \nabla_y G_m, \nabla_y H_1, \dots, \nabla_y H_p \right]^T. \quad (5)$$

At a Kuhn-Tucker point, (3b) could be written as

$$\bar{R}[X(y), y] = 0. \quad (3c)$$

Totally differentiating (3c) with respect to  $y$ , the following relationship is obtained.

$$J_X \bar{R}(X(y), y) \frac{dX(y)}{dy} + J_y \bar{R}(X(y), y) = 0. \quad (6)$$

The principal result immediately follows.

$$\frac{dX(y)}{dy} = - J_X \bar{R}(X(y), y)^{-1} \cdot J_y \bar{R}(X(y), y). \quad (7)$$



Equation (7) is almost the same as the result stated under Theorem 1 but is stated with respect to the entire Kuhn-Tucker triple,  $X$ . Note that (7) is stated for  $y$  in a neighborhood of  $y_0$ , whereas the result of Theorem 1 was only at  $y_0$ . This is possible due to the slightly stronger assumptions made by Fiacco and McCormick [3] and Fiacco [2].

The above result is the basis for the development of a methodology for sensitivity analysis in nonlinear programming. The computational technique used here is the penalty function method. The same result and technique were used by Armacost and Mylander [1] to compute (1), the principal difference in their methodology being that they approximate  $J_y \bar{R}$  by central differencing techniques, and compute all other derivatives manually. The relevant formulae for the penalty function implementation using the logarithmic barrier-quadratic loss penalty function of the SUMT algorithm [4] are discussed in Section 3.

### 3. Sensitivity Analysis With Symbolic Input

The factorable input technique and computer code to compute gradient vectors and Hessian matrices of problem functions was developed by McCormick [5] to facilitate the use of nonlinear programming computer codes. The code itself has since been extended by de Silva and McCormick to accept input in a symbolic format. This symbolic code has been used to solve several general nonlinear programming problems and the ease with which a problem input can be supplied or modifications done is very encouraging. A sensitivity analysis option turned out to be a natural extension to the symbolic computer code.

A reduced version of the matrix in (1) which provides substantial computational economy, results from the formulae which are developed below; a brief discussion of the factorable input philosophy precedes this development.

In the factorable input method, the problem functions are built up recursively as sums, or sums of pairwise products, of simple transformations of previously defined functions. Initially, these simple transformations

are applied to the original problem variables  $x_i$ ,  $i = 1, \dots, n$ , and the recursive application of the above procedure (shown mathematically in Equation (8)) generates new functions  $X^i$ ,  $i = n+1, \dots, N$ .

McCormick [5] gives a detailed exposition illustrated by an example. He sets  $X^i = x_i$ ,  $i = 1, \dots, n$  and refers to the entire set of  $X^i$ ,  $i = 1, \dots, N$  as concomitant variable functions (abbrev. CVFs). The general equation for the creation of a new CVF is:

$$X^i(x) = \sum_{p=1}^{i-1} T_p^i[X^p(x)] + \sum_{p=1}^{i-1} \sum_{q=1}^p V_{q,p}^i[X^p(x)] \cdot U_{p,q}^i[X^q(x)] . \quad (8)$$

In the presence of a parametric vector  $y$ , the equation becomes

$$X^i(x,y) = \sum_{p=1}^{i-1} T_p^i[X^p(x,y),y] + \sum_{p=1}^{i-1} \sum_{q=1}^p V_{q,p}^i[X^p(x,y),y] \cdot U_{p,q}^i[X^q(x,y),y] . \quad (9)$$

The terms in the first summation of (8) or (9) are referred to as separable terms comprising the new CVF  $X^i(\bullet)$  and the terms of the second summation of pairwise products are called quadratic terms. A point of note in (9) is that the arguments of the simple transformation

$T_p^i$ ,  $V_{q,p}^i$  or  $U_{p,q}^i$  contain  $y$  directly as well as indirectly via the

argument CVF. The direct presence of the parametric vector  $y$  is restricted to, at most, two components because all the simple transformations used in the methodology and code contain, at most, two constant terms. The parametric components are a subset of these constants.

An examination of (7) suggests that second partial derivatives of the problems functions with respect to  $x$  and the cross second partial derivatives with respect to  $x$  and  $y$  are required. This is because  $\bar{R}$  already contains first partial derivatives with respect to  $x$ . What is developed below will yield the matrix in (1) reduced by postmultiplying it

with a known column vector  $\beta \in E^k$  or premultiplying it with a known row vector  $\sigma \in E^n$ . The postmultiplication produces an n-vector, each component being a weighted sum of the k partial derivatives of a particular  $x_i(y)$ , with respect to the components of the parametric vector.

The weights are the components of  $\beta \in E^k$ . The ith component of this resulting n-vector can be looked upon as the "total sensitivity" of the variable  $x_i(y)$  ( $i=1, \dots, n$ ) with respect to all the  $y_j$ ,  $j = 1, \dots, k$ , weighted according to  $\beta_j$ ,  $j = 1, \dots, k$ . The rth component of the

n-vector is  $\sum_{j=1}^k \frac{\partial x_r}{\partial y_j} \beta_j$ . The premultiplication produces a k-component

vector, the jth component being the "total sensitivity" of a composite of all n variables  $x_i(y)$ ,  $i = 1, \dots, n$ , weighted according to  $\sigma_i$ ,  $i = 1, \dots, n$ , with respect to  $y_j$  ( $j=1, \dots, k$ ). The lth component of this k-vector

would be  $\sum_{i=1}^n \frac{\partial x_i}{\partial y_l} \sigma_i$ .

For clarity of exposition, (9) will be simplified below to a separable portion of one term or a quadratic term of one product.

Consider the separable case first. Suppose

$$X^i(x, y) = T[X^P(x, y), y] \quad (10)$$

Letting

$$\dot{T} \triangleq \dot{T}[X^P(x, y), y] \triangleq \frac{\partial T[X^P(x, y), y]}{\partial X^P} \quad (11)$$

we have

$$\nabla_x X^i(x, y) = \nabla_x X^P(x, y) \cdot \frac{\partial T[X^P(x, y), y]}{\partial X^P} \quad (12a)$$

$$= \dot{T} \cdot \nabla_x X^P(x, y) \quad (12b)$$



In (12b)  $\dot{T}$  is readily available since all transformations are simple ones, and  $\nabla_x X^P(x,y)$  is built up inductively from its argument CVF.

The expressions for  $\nabla_{xx}^2 X^i(x,y)$  are already developed and coded in [5], and will not be repeated here. Note that

$$\nabla_{yx}^2 X^i(x,y) = \nabla_y \nabla_x X^i(x,y) \quad (13a)$$

$$= D_y D_x T[X^P(x,y),y] \quad (13b)$$

In (13b)  $D$  denotes total differentiation, although in the previous step the partial differentiation  $\nabla$  was used. This is due to the fact that the partial differentiation in (13a) would capture the total variation due to  $x$  and  $y$  of the expression in (12a), but not so in (13b). Continuing the differentiation with respect to  $y$  of (12b), as in (13b), we get

$$\begin{aligned} \nabla_{yx}^2 X^i(x,y) &= \nabla_{yx}^2 X^P(x,y) \cdot \dot{T} + \nabla_x X^P(x,y) \left\{ \ddot{T} [X^P(x,y),y] \nabla_y^T X^P(x,y) \right. \\ &\quad \left. + \nabla_y^T \dot{T} \right\}. \end{aligned} \quad (14)$$

$$\text{In the above, } \ddot{T} \triangleq \ddot{T} [X^P(x,y),y] \triangleq \frac{\partial^2 T[X^P(x,y),y]}{\partial (X^P)^2}. \quad (15)$$

In (14),  $\dot{T}$  and  $\ddot{T}$  are readily available scalar quantities. The quantity  $\nabla_y^T \dot{T} [X^P(x,y),y]$  is the true partial derivative unlike in (13b). This is easy to compute because  $y$  enters the  $\dot{T}$  argument directly, at most, in two nonzero components. The expression for  $\nabla_x X^P(x,y)$  has been derived already in (12a). The only remaining terms are  $\nabla_{yx}^2 X^P(x,y)$  which is built up inductively, and  $\nabla_y^T X^P(x,y)$  which is discussed below.



Now

$$\begin{aligned} \nabla_y X^i(x,y) &= D_y T[X^p(x,y), y] \\ &= \nabla_y T[X^p(x,y), y] + \dot{T} \cdot \nabla_y X^p(x,y) . \end{aligned} \quad (16)$$

Note  $\nabla_y T(X^p(x,y), y)$  is the true partial derivative and will have at most two nonzero components in  $E^k$  which are easily computable. Thus (16) permits the computation of the  $\nabla_y^T X^p(x,y)$  term of (14) by induction. Since the formulae of (12b), (14), (16) are initially applied to the original problem variables  $X_i = x^i$ ,  $i = 1, \dots, n$ , we have, for  $p \leq n$ :

$$\nabla_x X^i(x,y) = e_p \cdot \dot{T} \quad (17)$$

where  $e_j$  is a unit vector  $\in E^n$  with unity in the  $j$ th position.

$$\begin{aligned} \text{Also,} \quad \nabla_{yx}^2 X^i(x,y) &= \nabla_{yx}^2 X^p + e_p [\dot{T} \nabla^T X^p + \nabla_y^T \dot{T}] \\ &= e_p \nabla_y^T \dot{T} . \end{aligned} \quad (18)$$

Finally,

$$\begin{aligned} \nabla_y X^i(x,y) &= \nabla_y T + \nabla_y X^p \dot{T} \\ &= \nabla_y T . \end{aligned} \quad (19)$$

On page 6, we spoke of dimensionally reducing the matrix of (1) in two separate ways. The case of premultiplying by  $\alpha \in E^n$  is considered below.

Thus, from (17) and (18) letting  $\alpha_p$  denote the  $p$ th component of  $\alpha$ , we get

$$\alpha^T \nabla_x X^i(x,y) = \alpha_p \cdot \dot{T} . \quad (20)$$

$$\alpha^T \nabla_{yx}^2 X^i(x,y) = \alpha_p \nabla_y^T \dot{T} . \quad (21)$$

The summation of (21) will have, at most, two nonzero components due to the nature of the simple transformations used. Next, consider the case of  $p > n$ . Here,

$$\alpha^T \nabla_x X^i(x,y) = \alpha^T \nabla_x X^p(x,y) \cdot \dot{T} . \quad (22)$$

Next,

$$\begin{aligned} \alpha^T \nabla_{yx}^2 X^i(x,y) &= \alpha^T \nabla_{yx}^2 X^p(x,y) \cdot \dot{T} \\ &+ \alpha^T \nabla_x X^p(x,y) \left\{ \ddot{T} \nabla_y^T X^p(x,y) + \nabla_y^T \dot{T} \right\} . \end{aligned} \quad (23)$$

Equation (22) works purely inductively and does not require the deliberate introduction of  $\alpha$ , as this has been done in (20). It is a scalar quantity for each  $i$ . The first term on the right-hand side of (23) is also an inductive term. As far as the second term is concerned, (the scalar multiplier  $\alpha^T \nabla_x X^p(x,y)$ , the only presence of  $\alpha$  there,) it is built up inductively. The expression within the curly bracket is a k-vector, the first term being an inductive one as demonstrated in (16) and the second possessing, at most, two nonzero components.

In summary, we have a scalar quantity for the reduced version of the gradient vector of each CVF  $i$ , and (20) (or (17), if  $p \leq n$ ) has to be repeated for every  $p$  corresponding to the separable terms of  $X^i$ , and added together. The reduced second cross partial of each CVF  $i$  (as shown in (21) or (18) depending on whether  $p \leq n$ ) is a k-vector for each component  $p$ , which must be added together as before.

The quadratic components of (9) will be addressed next. Considering the creation of an  $X^i$ ,  $i = n+1, \dots, N$ , using a single product, we have

$$X^i(x,y) = V[X^p(x,y), y] * U[X^q(x,y), y] , \quad (24)$$

then,

$$\begin{aligned} \nabla_x X^i(x,y) &= D_x V[X^p(x,y), y] * U[X^q(x,y), y] \\ &+ V[X^p(x,y), y] * D_x U[X^q(x,y), y] \end{aligned} \quad (25)$$

$$= \dot{V}U \cdot \nabla_x X^p(x,y) + V\dot{U} \cdot \nabla_x X^q(x,y) . \quad (26)$$

Now  $\dot{V}$ ,  $\dot{U}$  are defined analogously to  $\dot{T}$  (see (11)). Since  $V$ ,  $\dot{V}$ ,  $U$ ,  $\dot{U}$  are scalar quantities, the premultiplication by  $\alpha \in E^n$  could be done as in (20) and (22) depending on whether  $p$  and  $q$  are  $\leq$  or  $> n$ . The separate terms in (25) must be added together. The result is

$$\alpha^T \nabla_x^2 X^i(x, y) = \dot{V} U \alpha^T \nabla_x^2 X^p(x, y) + \dot{U} V \alpha^T \nabla_x^2 X^q(x, y). \quad (26)$$

When  $p$  (or  $q$ ) is  $> n$ ,  $\alpha^T \nabla_x^2 X^p$  (or  $\alpha^T \nabla_x^2 X^q$ ) enters purely inductively. Otherwise, the term simplifies to  $\dot{V} U \alpha_p$  (or  $\dot{U} V \alpha_q$ ). The second cross partial derivative is obtained by differentiating (26) with respect to  $y$ . The arguments of  $X^p$  and  $X^q$  are omitted for convenience of representation. Finally,

$$\begin{aligned} \nabla_{yx}^2 X^i(x, y) = U \bigg\{ & \dot{V} \cdot \nabla_{yx}^2 X^p + \nabla_x X^p [\dot{V} \cdot \nabla_y^T X^p + \nabla_y^T \dot{V}] \bigg\} \\ & + \dot{V} \cdot \nabla_x X^p \cdot D_y U \\ & + V \bigg\{ \dot{U} \cdot \nabla_{yx}^2 X^q + \nabla_x X^q [\dot{U} \cdot \nabla_y^T X^q + \nabla_y^T \dot{U}] \bigg\} \\ & + \dot{U} \cdot \nabla_x X^q \cdot D_y V. \end{aligned} \quad (27)$$

Premultiplying (27) by  $\alpha \in E^n$ , and rearranging,

$$\begin{aligned} \alpha^T \nabla_{yx}^2 X^i(x, y) = U \bigg\{ & \alpha^T \nabla_{yx}^2 X^p \cdot \dot{V} + \alpha^T \nabla_x X^p [\dot{V} \cdot \nabla_y^T X^p + \nabla_y^T \dot{V}] \bigg\} \\ & + V \bigg\{ \alpha^T \nabla_{yx}^2 X^q \cdot \dot{U} + \alpha^T \nabla_x X^q [\dot{U} \cdot \nabla_y^T X^q + \nabla_y^T \dot{U}] \bigg\} \\ & + \alpha^T \nabla_x X^p \cdot D_y U \cdot \dot{V} + \alpha^T \nabla_x X^q \cdot \nabla_y D \cdot \dot{U}. \end{aligned} \quad (28)$$

The evaluation of the portions within the curly brackets in (28) is identical to the evaluation of the cross second partials in the separable analysis, i.e., the right-hand side of (23). The last two terms of (26) contain two quantities  $\alpha^T \nabla_x^2 X^p$  and  $\alpha^T \nabla_x^2 X^q$  which are developed by induction as demonstrated by (12) and (20). The quantities  $D_y U$  and  $D_y V$



are analogous to  $D_y^T$  whose computation is shown in (16). The only remaining expression to be developed is that for  $\nabla_y X^1$ ,

$$\begin{aligned}\nabla_y X^1(x,y) &= D_y \{V[X^p(x,y),y] U[X^q(x,y),y]\} \\ &= U\{\dot{V} \cdot \nabla_y X^p + \nabla_y V\} + V\{\dot{U} \cdot \nabla_y X^q + \nabla_y U\}.\end{aligned}\quad (29)$$

Examination of (23) and (28) shows that it is not necessary to premultiply the expressions of (16) and (29) by  $\alpha^T$ , since  $\nabla_y X$  never gets directly premultiplied by it.

The above Equations (20) through (29) have developed the necessary components to perform the computations necessary to produce a row k-vector  $\alpha^T [dx(y)/dy]$ . The next section describes how these are implemented via a penalty function method.

#### 4. Computation of Sensitivity Results

The SUMT computer code [4] is used in conjunction with the factorable symbolic sensitivity analysis code to yield  $[\alpha^T dx(y)/dy]$ . The code uses an interior-exterior penalty function with a single parameter  $r$ , and generates a sequence of subproblems which will be denoted as  $M(y,r)$ , which tend to  $M(y)$  as  $r \rightarrow 0$ . In straight nonlinear optimization SUMT uses an extrapolation based on  $r$  to estimate the relevant values at  $r = 0$ . A similar extrapolation technique is used here to provide first order estimates of  $[\alpha^T dx(y)/dy]$ .

A logarithmic barrier quadratic loss penalty function is considered below. For  $M(y)$ , this penalty function is

$$P[x,y,r] = F(x,y) - r \sum_{i=1}^m \ln G_i(x,y) + \sum_{j=1}^p \frac{H_j^2(x,y)}{r}.\quad (30)$$

The problem solved at the  $l$ th stage of the sequential procedure, for a value  $r = r^l$  is

$$\begin{array}{ll}\text{MIN}_{x \in R^0} & P[x,y,r^l] \\ & \text{Problem } M(r,y)\end{array}$$

where

$$R^0 \equiv \{x | G_i(x,y) > 0, i=1, \dots, m\}.$$



The optimality condition is that the gradient with respect to  $x$  vanishes, i.e.,

$$\nabla_x P[x, y, r^t] = 0. \quad (31)$$

At local optima, for a general value of  $r$  we can write

$$\nabla_x P[x(y, r), y, r] \equiv 0. \quad (32)$$

Applying this to (30),

$$\begin{aligned} \nabla_x F[x(y, r), y] - \sum_{i=1}^m \frac{r}{G_i[x(y, r), y]} \cdot \nabla_x G_i[x(y, r), y] \\ + \sum_{j=1}^p \frac{2 H_j[x(y, r), y]}{r} \nabla_x H_j[x(y, r), y] \equiv 0. \end{aligned} \quad (33)$$

Differentiating (33) partially with respect to  $y$ , we get

$$\begin{aligned} \nabla_{yx}^2 P[x(y, r), y, r] \\ = \nabla_{yx}^2 F[x(y, r), y] - \sum_{i=1}^m \nabla_{yx}^2 G_i[x(y, r), y] \cdot \frac{r}{G_i[x(y, r), y]} \\ + \sum_{i=1}^m \frac{r}{G_i[x(y, r), y]^2} \cdot \nabla_x G_i[x(y, r), y] \cdot \nabla_y^T G_i[x(y, r), y] \\ + \sum_{j=1}^p \nabla_{yx}^2 H_j[x(y, r), y] \cdot \frac{2 H_j[x(y, r), y]}{r} \\ + \sum_{j=1}^p \left( \frac{2}{r} \right) \nabla_x H_j[x(y, r), y] \cdot \nabla_y^T H_j[x(y, r), y]. \end{aligned} \quad (34)$$

Differentiating (33) totally with respect to  $y$ ,

$$\nabla_{xx}^2 P[x(y, r), y, r] \frac{dx(y, r)}{dy} + \nabla_{yx}^2 P[x(y, r), y, r] \equiv 0. \quad (35)$$

The argument variables within the square brackets will be omitted when referring to the terms in the above.

$\nabla_{xx}^2 P$  has been developed elsewhere by McCormick [5]. Those for  $\nabla_{xy}^2 P$  are developed using the Equation (34). Thus,

$$\frac{d(x(y,r))}{dy} = - \nabla_{xx}^2 P[x(y,r), y, r]^{-1} \cdot \nabla_{yx}^2 P[x(y,r), y, r] . \quad (36)$$

The similarity between (7) and (36) is to be noted. Suppose (36) is premultiplied by some  $\sigma \in E^n$ . Then, we get

$$\sigma^T \frac{dx(y,r)}{dy} = - \alpha^T \cdot \nabla_{yx}^2 P[x(y,r), y, r] , \quad (37)$$

where

$$\alpha^T = \sigma^T \cdot \nabla_{xx}^2 P[x(y,r), y, r]^{-1} . \quad (38)$$

If  $\sigma$  is a known vector,  $\alpha$  is determined because the Hessian of the penalty function with respect to  $x$  is known by [5]. Then (37) is determined by premultiplying (34) by this  $\alpha$ , and using (20), (21), (22), (23), (27) and (29) to determine the individual components of the resulting equation. This can be done because every problem function  $F$ ,  $G_i$ ;  $i = 1, \dots, m$ ,  $H_j$ ;  $j = 1, \dots, p$ , can be identified with a CVF  $X^i$ ,  $i = 1, \dots, n$ ,  $n+1, \dots, N$ .

The final aspect remaining to be discussed in the context of the sensitivity methodology is the improvement of the estimate of  $\sigma^T \frac{dx(y,r)}{dy}$  by extrapolating on  $r$ , at the  $\ell$ th sequential unconstrained problem, to  $r = 0$ . A Taylor series approximation applied at the  $\ell$ th and  $(\ell-1)$ th subproblems yields

$$\frac{dx(y, r^{\ell})}{dy} = \frac{dx(y, 0)}{dy} + r^{\ell} \frac{d}{dr} \left[ \frac{dx(y, 0)}{dy} \right]. \quad (39)$$

$$\frac{dx(y, r^{\ell-1})}{dy} = \frac{dx(y, 0)}{dy} + r^{\ell-1} \frac{d}{dr} \left[ \frac{dx(y, 0)}{dy} \right]. \quad (40)$$

Eliminating the last term in (39) and (40)

$$\frac{d(x(y, 0))}{dy} (r^{\ell-1} - r^{\ell}) = r^{\ell-1} \cdot \frac{dx(y, r^{\ell})}{dy} - r^{\ell} \cdot \frac{dx(y, r^{\ell-1})}{dy} \quad (41)$$

from which, (letting  $r^{\ell-1} = c \cdot r^{\ell}$ ) we get

$$\frac{dx(y, 0)}{dy} = \frac{1}{c-1} \cdot \left[ c \frac{dx(y_1 r^{\ell})}{dy} - \frac{dx(y_1 r^{\ell-1})}{dy} \right]. \quad (42)$$

Hence by performing the computation for successive subproblems when  $r$  becomes sufficiently small, (42) can be used to estimate the value for the original problem, i.e.,  $r = 0$ .

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